# STK2130 - Lecture 13 

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## A continuous-time Markov chain

We recall the following from Chapter 6.2:
A continuous-time Markov chain with stationary transition probabilities and state space $\mathcal{X}$ is a stochastic process such that:

- The times spent in the different states are independent random variables (because of the Markov property).
- The amount of time spent in state $i \in \mathcal{X}$ is exponentially distributed with rate $v_{i}$ (because of the Markov property and stationarity).
- When the process leaves state $i$, it enters state $j$ with some transition probability $Q_{i j}$ where:

$$
\begin{aligned}
Q_{i i}=0, & \text { for all } i \in \mathcal{X} \\
\sum_{j \in \mathcal{X}} Q_{i j}=1, & \text { for all } i \in \mathcal{X}
\end{aligned}
$$

- The transitions follow a discrete-time Markov chain.


## A continuous-time Markov chain (cont.)

We now introduce the following notation:

$$
q_{i j}=v_{i} Q_{i j}, \quad i, j \in \mathcal{X}
$$

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INTERPRETATION: Since $v_{i}$ is the rate at which the process makes a transition when in state $i$ and $Q_{i j}$ is the probability that this transition is into state $j$, it follows that $q_{i j}$ is the rate, when in state $i$, at which the process makes a transition into state $j$.

The quantities $q_{i j}$ are called the instantaneous transition rates.

## A continuous-time Markov chain (cont.)

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The quantities $q_{i j}$ are called the instantaneous transition rates.
Since we have:

$$
\begin{aligned}
& v_{i}=v_{i} \sum_{j \in \mathcal{X}} Q_{i j}=\sum_{j \in \mathcal{X}} v_{i} Q_{i j}=\sum_{j \in \mathcal{X}} q_{i j}, \\
& Q_{i j}=\frac{q_{i j}}{v_{i}}=\frac{q_{i j}}{\sum_{j \in \mathcal{X}} q_{i j}},
\end{aligned}
$$

the probabilistic properties of $\{X(t): t \geq 0\}$ is determined by the $q_{i j}$ 's.

### 6.4 Kolmogorov's Backward Equations

Lemma (6.2)
Let $P_{i j}(t)=P(X(t)=j \mid X(0)=i), i, j \in \mathcal{X}$. We then have:
(a) $\lim _{h \rightarrow 0} \frac{1-P_{i i}(h)}{h}=v_{i}$
(b) $\lim _{h \rightarrow 0} \frac{P_{i j}(h)}{h}=q_{i j}, \quad$ for all $i \neq j$

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PROOF: We start by noting that the amount of time until a transition occurs is exponentially distributed.

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PROOF: We start by noting that the amount of time until a transition occurs is exponentially distributed.

Hence, the probability of two or more transitions in a time $h$ is $o(h)$.

### 6.4 Kolmogorov's Backward Equations (cont.)

Hence, it follows that:

$$
P(X(h) \neq i \mid X(0)=i)=1-P(X(h)=i \mid X(0)=i)=1-P_{i i}(h)=v_{i} h+o(h) .
$$

### 6.4 Kolmogorov's Backward Equations (cont.)

Hence, it follows that:

$$
P(X(h) \neq i \mid X(0)=i)=1-P(X(h)=i \mid X(0)=i)=1-P_{i i}(h)=v_{i} h+o(h) .
$$

By dividing both sides by $h$ and letting $h$ go to 0 , we get:

$$
\lim _{h \rightarrow 0} \frac{1-P_{i i}(h)}{h}=\lim _{h \rightarrow 0} \frac{v_{i} h+o(h)}{h}=v_{i}+\lim _{h \rightarrow 0} \frac{o(h)}{h}=v_{i}
$$

which proves (a).

### 6.4 Kolmogorov's Backward Equations (cont.)

Similarly, if $i \neq j$, we get:

$$
P(X(h)=j \mid X(0)=i)=P_{i j}(h)=v_{i} Q_{i j} h+o(h) .
$$

### 6.4 Kolmogorov's Backward Equations (cont.)

Similarly, if $i \neq j$, we get:

$$
P(X(h)=j \mid X(0)=i)=P_{i j}(h)=v_{i} Q_{i j} h+o(h) .
$$

By dividing both sides by $h$ and letting $h$ go to 0 , we get:

$$
\lim _{h \rightarrow 0} \frac{P_{i j}(h)}{h}=\lim _{h \rightarrow 0} \frac{v_{i} Q_{i j} h+o(h)}{h}=v_{i} Q_{i j}+\lim _{h \rightarrow 0} \frac{o(h)}{h}=v_{i} Q_{i j}=q_{i j}
$$

which proves (b).

### 6.4 Kolmogorov’s Backward Equations (cont.)

Lemma (6.3 - Chapman-Kolmogorov equations)
For all $s, t \geq 0$ and $i, j \in \mathcal{X}$ we have:

$$
P_{i j}(t+s)=\sum_{k \in \mathcal{X}} P_{i k}(t) P_{k j}(s)
$$

### 6.4 Kolmogorov’s Backward Equations (cont.)

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$$
P_{i j}(t+s)=\sum_{k \in \mathcal{X}} P_{i k}(t) P_{k j}(s)
$$

PROOF: We have:

$$
\begin{aligned}
P_{i j}(t+s) & =P(X(t+s)=j \mid X(0)=i) \\
& =\sum_{k \in \mathcal{X}} P(X(t+s)=j, X(t)=k \mid X(0)=i)
\end{aligned}
$$

### 6.4 Kolmogorov’s Backward Equations (cont.)

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& =\sum_{k \in \mathcal{X}} P(X(t+s)=j, X(t)=k \mid X(0)=i) \\
& =\sum_{k \in \mathcal{X}} P(X(t+s)=j \mid X(t)=k, X(0)=i) \cdot P(X(t)=k \mid X(0)=i)
\end{aligned}
$$

### 6.4 Kolmogorov’s Backward Equations (cont.)

$$
P_{i j}(t+s)=\sum_{k \in \mathcal{X}} P(X(t+s)=j \mid X(t)=k, X(0)=i) \cdot P(X(t)=k \mid X(0)=i)
$$

### 6.4 Kolmogorov’s Backward Equations (cont.)

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\begin{aligned}
P_{i j}(t+s) & =\sum_{k \in \mathcal{X}} P(X(t+s)=j \mid X(t)=k, X(0)=i) \cdot P(X(t)=k \mid X(0)=i) \\
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\end{aligned}
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P_{i j}(t+s) & =\sum_{k \in \mathcal{X}} P(X(t+s)=j \mid X(t)=k, X(0)=i) \cdot P(X(t)=k \mid X(0)=i) \\
& =\sum_{k \in \mathcal{X}} P(X(t+s)=j \mid X(t)=k) \cdot P(X(t)=k \mid X(0)=i) \\
& =\sum_{k \in \mathcal{X}} P(X(s)=j \mid X(0)=k) \cdot P(X(t)=k \mid X(0)=i) \\
& =\sum_{k \in \mathcal{X}} P_{k j}(s) \cdot P_{i k}(t)
\end{aligned}
$$

### 6.4 Kolmogorov's Backward Equations (cont.)

Theorem (6.1 - Kolmogorov's backward equations)
For all $t \geq 0$ and states $i, j \in \mathcal{X}$ we have:

$$
P_{i j}^{\prime}(t)=\sum_{k \in \mathcal{X} \backslash i} q_{i k} P_{k j}(t)-v_{i} P_{i j}(t) .
$$

### 6.4 Kolmogorov's Backward Equations (cont.)

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For all $t \geq 0$ and states $i, j \in \mathcal{X}$ we have:

$$
P_{i j}^{\prime}(t)=\sum_{k \in \mathcal{X} \backslash i} q_{i k} P_{k j}(t)-v_{i} P_{i j}(t) .
$$

PROOF: By Lemma 6.3 we have:

$$
P_{i j}(t+h)-P_{i j}(t)=\sum_{k \in \mathcal{X}} P_{i k}(h) P_{k j}(t)-P_{i j}(t)
$$

### 6.4 Kolmogorov’s Backward Equations (cont.)

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For all $t \geq 0$ and states $i, j \in \mathcal{X}$ we have:

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$$

PROOF: By Lemma 6.3 we have:

$$
\begin{aligned}
P_{i j}(t+h)-P_{i j}(t) & =\sum_{k \in \mathcal{X}} P_{i k}(h) P_{k j}(t)-P_{i j}(t) \\
& =\sum_{k \in \mathcal{X} \backslash i} P_{i k}(h) P_{k j}(t)+P_{i j}(h) P_{i j}(t)-P_{i j}(t)
\end{aligned}
$$

### 6.4 Kolmogorov’s Backward Equations (cont.)

Theorem (6.1 - Kolmogorov's backward equations)
For all $t \geq 0$ and states $i, j \in \mathcal{X}$ we have:

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$$

PROOF: By Lemma 6.3 we have:

$$
\begin{aligned}
P_{i j}(t+h)-P_{i j}(t) & =\sum_{k \in \mathcal{X}} P_{i k}(h) P_{k j}(t)-P_{i j}(t) \\
& =\sum_{k \in \mathcal{X} \backslash i} P_{i k}(h) P_{k j}(t)+P_{i j}(h) P_{i j}(t)-P_{i j}(t) \\
& =\sum_{k \in \mathcal{X} \backslash i} P_{i k}(h) P_{k j}(t)-\left[1-P_{i i}(h)\right] P_{i j}(t)
\end{aligned}
$$

### 6.4 Kolmogorov's Backward Equations (cont.)

By dividing both sides by $h$ and letting $h \rightarrow 0$, we can use Lemma 6.2 and get:

$$
\begin{aligned}
\lim _{h \rightarrow 0} \frac{P_{i j}(t+h)-P_{i j}(t)}{h} & =\sum_{k \in \mathcal{X} \backslash i} \lim _{h \rightarrow 0} \frac{P_{i k}(h)}{h} P_{k j}(t)-\lim _{h \rightarrow 0} \frac{1-P_{i i}(h)}{h} P_{i j}(t) \\
& =\sum_{k \in \mathcal{X} \backslash i} q_{i k} P_{k j}(t)-v_{i} P_{i j}(t)
\end{aligned}
$$

### 6.4 Kolmogorov's Backward Equations (cont.)

Kolmogorov's backward equations:

$$
P_{i j}^{\prime}(t)=\sum_{k \in \mathcal{X} \backslash i} q_{i k} P_{k j}(t)-v_{i} P_{i j}(t), \quad i, j \in \mathcal{X} .
$$

We recall that $v_{i}=\sum_{k \in \mathcal{X}} q_{i, k}=\sum_{k \in \mathcal{X}} v_{i} Q_{i, k}$. Since $Q_{i, i}=0$, this implies that:

$$
v_{i}=\sum_{k \in \mathcal{X} \backslash i} q_{i, k}
$$

Hence, we have:

$$
\sum_{k \in \mathcal{X} \backslash i} q_{i, k}-v_{i}=\sum_{k \in \mathcal{X} \backslash i} q_{i, k}-\sum_{k \in \mathcal{X} \backslash i} q_{i, k}=0
$$

Thus, in each of the backward equations the sum of coefficients is equal to zero.

### 6.4 Kolmogorov’s Backward Equations (cont.)

We now assume that $\mathcal{X}=\{1,2, \ldots, n\}$, and introduce the following matrices:

$$
\begin{gathered}
\boldsymbol{R}=\left[\begin{array}{ccccc}
-v_{1} & q_{1,2} & q_{1,3} & \cdots & q_{1, n} \\
q_{2,1} & -v_{2} & q_{2,3} & \cdots & q_{2, n} \\
q_{3,1} & q_{3,2} & -v_{3} & \cdots & q_{3, n} \\
\vdots & \vdots & \vdots & \ddots & \vdots \\
q_{n, 1} & q_{n, 2} & q_{n, 3} & \cdots & -v_{n}
\end{array}\right] \\
\boldsymbol{P}(t)=\left[\begin{array}{ccccc}
P_{1,1}(t) & P_{1,2}(t) & P_{1,3}(t) & \cdots & P_{1, n}(t) \\
P_{2,1}(t) & P_{2,2}(t) & P_{2,3}(t) & \cdots & P_{2, n}(t) \\
P_{3,1}(t) & P_{3,2}(t) & P_{3,3}(t) & \cdots & P_{3, n}(t) \\
\vdots & \vdots & \vdots & \ddots & \vdots \\
P_{n, 1}(t) & P_{n, 2}(t) & P_{n, 3}(t) & \cdots & P_{n, n}(t)
\end{array}\right]
\end{gathered}
$$

### 6.4 Kolmogorov's Backward Equations (cont.)

It is then easy to see that Kolmogorov's backward equations:

$$
P_{i j}^{\prime}(t)=\sum_{k \in \mathcal{X} \backslash i} q_{i k} P_{k j}(t)-v_{i} P_{i j}(t) .
$$

can be written in the following form:

$$
\boldsymbol{P}^{\prime}(t)=\boldsymbol{R} \boldsymbol{P}(t)
$$

## Example 6.10 - Pure birth process

For a pure birth process with birth rates $\lambda_{0}, \lambda_{1}, \ldots$ we have:

$$
\begin{aligned}
q_{i, i+1} & =\lambda_{i}, \quad i=0,1, \ldots \\
q_{i, j} & =0, \quad \text { for all } j \neq(i+1)
\end{aligned}
$$

Hence, we also have:

$$
v_{i}=\sum_{j=0}^{\infty} q_{i j}=q_{i, i+1}=\lambda_{i}, \quad i=0,1, \ldots
$$

By inserting this into the backward equations we get:

$$
\begin{aligned}
P_{i j}^{\prime}(t) & =\sum_{k \in \mathcal{X} \backslash i} q_{i k} P_{k j}(t)-v_{i} P_{i j}(t) \\
& =\lambda_{i} P_{i+1, j}(t)-\lambda_{i} P_{i j}(t), \quad t \geq 0, \quad 0 \leq i \leq j
\end{aligned}
$$

## Example 6.10 - Birth and death process

For a birth and death process with birth rates $\lambda_{0}, \lambda_{1}, \ldots$ and death rates $\mu_{1}, \mu_{2}, \ldots$, we have:

$$
\begin{aligned}
q_{i, i+1} & =\lambda_{i}, \quad i=0,1, \ldots \\
q_{i, i-1} & =\mu_{i}, \quad i=1,2, \ldots \\
q_{i, j} & =0, \quad \text { otherwise }
\end{aligned}
$$

Hence, we also have:

$$
\begin{aligned}
v_{0} & =\sum_{j=0}^{\infty} q_{0 j}=\lambda_{0} \\
v_{i} & =\sum_{j=0}^{\infty} q_{i j}=\lambda_{i}+\mu_{i}, \quad i=1,2, \ldots
\end{aligned}
$$

## Example 6.10 - Birth and death process (cont.)

By inserting this into the backward equations we get:

$$
\begin{aligned}
P_{0 j}^{\prime}(t) & =\sum_{k \in \mathcal{X} \backslash 0} q_{0 k} P_{k j}(t)-v_{0} P_{0 j}(t) \\
& =\lambda_{0} P_{1, j}(t)-\lambda_{0} P_{0 j}(t), \quad t \geq 0, \quad j \geq 0 \\
P_{i j}^{\prime}(t) & =\sum_{k \in \mathcal{X} \backslash i} q_{i k} P_{k j}(t)-v_{i} P_{i j}(t) \\
& =\lambda_{i} P_{i+1, j}(t)+\mu_{i} P_{i-1, j}(t)-\left(\lambda_{i}+\mu_{i}\right) P_{i j}(t), \quad t \geq 0, \quad i>0, j \geq 0
\end{aligned}
$$

## Example 6.11

The lifetimes and repair times of a system are independent and exponentially distributed with rates respectively $\lambda$ and $\mu$.

We model this system as a continuous-time Markov chain $\{X(t): t \geq 0\}$ with state space $\mathcal{X}=\{0,1\}$, where ${ }^{1}$ :

$$
X(t)=I(\text { The system is functioning at time } t), \quad t \geq 0 .
$$

The only non-zero instantaneous transition rates are $q_{01}=\mu$ and $q_{10}=\lambda$. Hence;

$$
\begin{aligned}
& v_{0}=\sum_{j \in \mathcal{X}} q_{0 j}=q_{01}=\mu \\
& v_{1}=\sum_{j \in \mathcal{X}} q_{1 j}=q_{10}=\lambda
\end{aligned}
$$

[^0]
## Example 6.11 (cont.)

By inserting this into the backward equations we get:

$$
\begin{align*}
& P_{00}^{\prime}(t)=\mu P_{10}(t)-\mu P_{00}(t)=\mu\left[P_{10}(t)-P_{00}(t)\right],  \tag{1}\\
& P_{01}^{\prime}(t)=\mu P_{11}(t)-\mu P_{01}(t)=\mu\left[P_{11}(t)-P_{01}(t)\right],  \tag{2}\\
& P_{10}^{\prime}(t)=\lambda P_{00}(t)-\lambda P_{10}(t)=\lambda\left[P_{00}(t)-P_{10}(t)\right],  \tag{3}\\
& P_{11}^{\prime}(t)=\lambda P_{01}(t)-\lambda P_{11}(t)=\lambda\left[P_{01}(t)-P_{11}(t)\right] \tag{4}
\end{align*}
$$

We start by computing $P_{11}(t)$. In order to solve this, we multiply (2) by $\lambda$ and multiply (4) by $\mu$ and add the resulting equations:

$$
\mu P_{11}^{\prime}(t)+\lambda P_{01}^{\prime}(t)=0
$$

By integrating both sides we get:

$$
\mu P_{11}(t)+\lambda P_{01}(t)=c
$$

## Example 6.11 (cont.)

In order to determine the constant $c$, we note that $P_{11}(0)=1$, while $P_{01}(0)=0$. Hence, $c=\mu$, and we get:

$$
\mu P_{11}(t)+\lambda P_{01}(t)=\mu
$$

or equivalently:

$$
\lambda P_{01}(t)=\mu\left[1-P_{11}(t)\right]
$$

We insert this into the right-hand side of (4) $P_{11}^{\prime}(t)=\lambda P_{01}(t)-\lambda P_{11}(t)$ and get:

$$
\begin{aligned}
P_{11}^{\prime}(t) & =\mu\left[1-P_{11}(t)\right]-\lambda P_{11}(t) \\
& =\mu-(\mu+\lambda) P_{11}(t)
\end{aligned}
$$

or equivalently:

$$
P_{11}^{\prime}(t)+(\mu+\lambda) P_{11}(t)=\mu
$$

## Example 6.11 (cont.)

In order to solve the differential equation $P_{11}^{\prime}(t)+(\mu+\lambda) P_{11}(t)=\mu$, we multiply both sides by the integrating factor $e^{(\mu+\lambda) t}$ and get:

$$
P_{11}^{\prime}(t) e^{(\mu+\lambda) t}+(\mu+\lambda) e^{(\mu+\lambda) t} P_{11}(t)=\mu e^{(\mu+\lambda) t}
$$

## Example 6.11 (cont.)

In order to solve the differential equation $P_{11}^{\prime}(t)+(\mu+\lambda) P_{11}(t)=\mu$, we multiply both sides by the integrating factor $e^{(\mu+\lambda) t}$ and get:

$$
P_{11}^{\prime}(t) e^{(\mu+\lambda) t}+(\mu+\lambda) e^{(\mu+\lambda) t} P_{11}(t)=\mu e^{(\mu+\lambda) t}
$$

By using the product rule for derivatives, the left-hand side can be simplified to:

$$
\left(P_{11}(t) e^{(\mu+\lambda) t}\right)^{\prime}=\mu e^{(\mu+\lambda) t}
$$

## Example 6.11 (cont.)

In order to solve the differential equation $P_{11}^{\prime}(t)+(\mu+\lambda) P_{11}(t)=\mu$, we multiply both sides by the integrating factor $e^{(\mu+\lambda) t}$ and get:

$$
P_{11}^{\prime}(t) e^{(\mu+\lambda) t}+(\mu+\lambda) e^{(\mu+\lambda) t} P_{11}(t)=\mu e^{(\mu+\lambda) t}
$$

By using the product rule for derivatives, the left-hand side can be simplified to:

$$
\left(P_{11}(t) e^{(\mu+\lambda) t}\right)^{\prime}=\mu e^{(\mu+\lambda) t}
$$

By integrating both sides of this equation we get:

$$
P_{11}(t) e^{(\mu+\lambda) t}=\frac{\mu}{\mu+\lambda} e^{(\mu+\lambda) t}+C
$$

## Example 6.11 (cont.)

In order to solve the differential equation $P_{11}^{\prime}(t)+(\mu+\lambda) P_{11}(t)=\mu$, we multiply both sides by the integrating factor $e^{(\mu+\lambda) t}$ and get:

$$
P_{11}^{\prime}(t) e^{(\mu+\lambda) t}+(\mu+\lambda) e^{(\mu+\lambda) t} P_{11}(t)=\mu e^{(\mu+\lambda) t}
$$

By using the product rule for derivatives, the left-hand side can be simplified to:

$$
\left(P_{11}(t) e^{(\mu+\lambda) t}\right)^{\prime}=\mu e^{(\mu+\lambda) t}
$$

By integrating both sides of this equation we get:

$$
P_{11}(t) e^{(\mu+\lambda) t}=\frac{\mu}{\mu+\lambda} e^{(\mu+\lambda) t}+C
$$

or equivalently:

$$
P_{11}(t)=\frac{\mu}{\mu+\lambda}+C e^{-(\mu+\lambda) t}
$$

## Example 6.11 (cont.)

In order to determine the constant $C$, we again use that $P_{11}(0)=1$. That is:

$$
1=\frac{\mu}{\mu+\lambda}+C
$$

Hence, $C$ is given by:

$$
C=1-\frac{\mu}{\mu+\lambda}=\frac{\lambda}{\mu+\lambda},
$$

and thus:

$$
P_{11}(t)=\frac{\mu}{\mu+\lambda}+\frac{\lambda}{\mu+\lambda} e^{-(\mu+\lambda) t}
$$

## Example 6.11 (cont.)

This also implies that:

$$
\begin{aligned}
P_{10}(t) & =1-P_{11}(t) \\
& =1-\frac{\mu}{\mu+\lambda}-\frac{\lambda}{\mu+\lambda} e^{-(\mu+\lambda) t} \\
& =\frac{\lambda}{\mu+\lambda}-\frac{\lambda}{\mu+\lambda} e^{-(\mu+\lambda) t}
\end{aligned}
$$

We recall that we also have established that $\mu P_{11}(t)+\lambda P_{01}(t)=\mu$. Hence, we get that:

$$
\begin{aligned}
P_{01}(t) & =\frac{\mu}{\lambda}\left[1-P_{11}(t)\right]=\frac{\mu}{\lambda} P_{10}(t) \\
& =\frac{\mu}{\mu+\lambda}-\frac{\mu}{\mu+\lambda} e^{-(\mu+\lambda) t}
\end{aligned}
$$

## Example 6.11 (cont.)

Finally, we get that:

$$
\begin{aligned}
P_{00}(t) & =1-P_{01}(t) \\
& =1-\frac{\mu}{\mu+\lambda}+\frac{\mu}{\mu+\lambda} e^{-(\mu+\lambda) t} \\
& =\frac{\lambda}{\mu+\lambda}+\frac{\mu}{\mu+\lambda} e^{-(\mu+\lambda) t}
\end{aligned}
$$

## Example 6.11 (cont.)

Summarizing all these results we get:

$$
\begin{aligned}
& P_{11}(t)=\frac{\mu}{\mu+\lambda}+\frac{\lambda}{\mu+\lambda} e^{-(\mu+\lambda) t} \\
& P_{10}(t)=\frac{\lambda}{\mu+\lambda}-\frac{\lambda}{\mu+\lambda} e^{-(\mu+\lambda) t} \\
& P_{01}(t)=\frac{\mu}{\mu+\lambda}-\frac{\mu}{\mu+\lambda} e^{-(\mu+\lambda) t} \\
& P_{00}(t)=\frac{\lambda}{\mu+\lambda}+\frac{\mu}{\mu+\lambda} e^{-(\mu+\lambda) t}
\end{aligned}
$$

## Example 6.11 (cont.)

We observe that:

$$
\lim _{t \rightarrow \infty} P_{11}(t)=\lim _{t \rightarrow \infty} P_{01}(t)=\frac{\mu}{\mu+\lambda}=\frac{\lambda^{-1}}{\mu^{-1}+\lambda^{-1}} .
$$

and that:

$$
\lim _{t \rightarrow \infty} P_{00}(t)=\lim _{t \rightarrow \infty} P_{10}(t)=\frac{\lambda}{\mu+\lambda}=\frac{\mu^{-1}}{\mu^{-1}+\lambda^{-1}} .
$$

### 6.4 Kolmogorov's Forward Equations

Theorem (6.2 - Kolmogorov's forward equations)
For all $t \geq 0$ and states $i, j \in \mathcal{X}$ we have:

$$
P_{i j}^{\prime}(t)=\sum_{k \in \mathcal{X} \backslash} P_{i k}(t) q_{k j}-P_{i j}(t) v_{j} .
$$

### 6.4 Kolmogorov's Forward Equations

Theorem (6.2 - Kolmogorov's forward equations)
For all $t \geq 0$ and states $i, j \in \mathcal{X}$ we have:

$$
P_{i j}^{\prime}(t)=\sum_{k \in \mathcal{X} \backslash} P_{i k}(t) q_{k j}-P_{i j}(t) v_{j} .
$$

PROOF: By Lemma 6.3 we have:

$$
P_{i j}(t+h)-P_{i j}(t)=\sum_{k \in \mathcal{X}} P_{i k}(t) P_{k j}(h)-P_{i j}(t)
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$$

PROOF: By Lemma 6.3 we have:

$$
\begin{aligned}
P_{i j}(t+h)-P_{i j}(t) & =\sum_{k \in \mathcal{X}} P_{i k}(t) P_{k j}(h)-P_{i j}(t) \\
& =\sum_{k \in \mathcal{X} \backslash} P_{i k}(t) P_{k j}(h)+P_{i j}(t) P_{i j}(h)-P_{i j}(t)
\end{aligned}
$$

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& =\sum_{k \in \mathcal{X} \backslash} P_{i k}(t) P_{k j}(h)+P_{i j}(t) P_{i j}(h)-P_{i j}(t) \\
& =\sum_{k \in \mathcal{X} \backslash} P_{i k}(t) P_{k j}(h)-P_{i j}(t)\left[1-P_{j j}(h)\right]
\end{aligned}
$$

### 6.4 Kolmogorov's Forward Equations (cont.)

By dividing both sides by $h$ and letting $h \rightarrow 0$, we can use Lemma $6.2^{2}$ and get:

$$
\begin{aligned}
\lim _{h \rightarrow 0} \frac{P_{i j}(t+h)-P_{i j}(t)}{h} & =\sum_{k \in \mathcal{X} \backslash j} P_{i k}(t) \lim _{h \rightarrow 0} \frac{P_{k j}(h)}{h}-P_{i j}(t) \lim _{h \rightarrow 0} \frac{1-P_{j j}(h)}{h} \\
& =\sum_{k \in \mathcal{X} \backslash j} P_{i k}(t) q_{k j}-P_{i j}(t) v_{j}
\end{aligned}
$$

${ }^{2}$ Unfortunately, the interchange of limit and summation is not always valid. This holds, however, for all birth and death processes and for all finite state models

### 6.4 Kolmogorov’s Forward Equations (cont.)

We again assume that $\mathcal{X}=\{1,2, \ldots, n\}$, and recall the following matrices:

$$
\begin{gathered}
\boldsymbol{R}=\left[\begin{array}{ccccc}
-v_{1} & q_{1,2} & q_{1,3} & \cdots & q_{1, n} \\
q_{2,1} & -v_{2} & q_{2,3} & \cdots & q_{2, n} \\
q_{3,1} & q_{3,2} & -v_{3} & \cdots & q_{3, n} \\
\vdots & \vdots & \vdots & \ddots & \vdots \\
q_{n, 1} & q_{n, 2} & q_{n, 3} & \cdots & -v_{n}
\end{array}\right] \\
\boldsymbol{P}(t)=\left[\begin{array}{ccccc}
P_{1,1}(t) & P_{1,2}(t) & P_{1,1}(t) & \cdots & P_{1, n}(t) \\
P_{2,1}(t) & P_{2,2}(t) & P_{2,3}(t) & \cdots & P_{2, n}(t) \\
P_{3,1}(t) & P_{3,2}(t) & P_{3,3}(t) & \cdots & P_{3, n}(t) \\
\vdots & \vdots & \vdots & \ddots & \vdots \\
P_{n, 1}(t) & P_{n, 2}(t) & P_{n, 3}(t) & \cdots & P_{n, n}(t)
\end{array}\right]
\end{gathered}
$$

### 6.4 Kolmogorov's Forward Equations (cont.)

It is then easy to see that Kolmogorov's forward equations:

$$
P_{i j}^{\prime}(t)=\sum_{k \in \mathcal{X} \backslash j} P_{i k}(t) q_{k j}-P_{i j}(t) v_{j}
$$

can be written in the following form:

$$
\boldsymbol{P}^{\prime}(t)=\boldsymbol{P}(t) \boldsymbol{R}
$$

## Pure birth processes

Assume that $\{X(t): t \geq 0\}$ is a pure birth process with birth rates $\lambda_{0}, \lambda_{1}, \ldots$. From this it follows that we have:

$$
\begin{aligned}
q_{i, i+1} & =\lambda_{i}, \quad i=0,1, \ldots \\
q_{i, j} & =0, \quad \text { for all } j \neq(i+1)
\end{aligned}
$$

Hence, we also have:

$$
v_{i}=\sum_{j=0}^{\infty} q_{i j}=q_{i, i+1}=\lambda_{i}, \quad i=0,1, \ldots
$$

By inserting this into the forward equations we get:

$$
\begin{aligned}
P_{i j}^{\prime}(t) & =\sum_{k \in \mathcal{X} \backslash} P_{i k}(t) q_{k j}-P_{i j}(t) v_{j} \\
& =P_{i, j-1}(t) \lambda_{j-1}-P_{i j}(t) \lambda_{j}, \quad t \geq 0, \quad i, j=0,1, \ldots
\end{aligned}
$$

## Pure birth processes (cont.)

Note, however, that $P_{i j}(t)=0$ whenever $j<i$ (since this is a pure birth process with no deaths).

Hence, we get the following simplified equations:

$$
\begin{aligned}
& P_{i i}^{\prime}(t)=\lambda_{i-1} P_{i, i-1}(t)-\lambda_{i} P_{i i}(t)=-\lambda_{i} P_{i i}(t), \quad i=0,1, \ldots \\
& P_{i j}^{\prime}(t)=\lambda_{j-1} P_{i, j-1}(t)-\lambda_{j} P_{i j}(t), \quad j \geq i+1
\end{aligned}
$$

## Pure birth processes (cont.)

Proposition (6.4 - Pure birth processes)
Assume that $\{X(t): t \geq 0\}$ is a pure birth process with birth rates $\lambda_{0}, \lambda_{1}, \ldots$. We then have:

$$
\begin{aligned}
& P_{i i}(t)=e^{-\lambda_{i} t}, \quad i=0,1,2, \ldots \\
& P_{i j}(t)=\lambda_{j-1} e^{-\lambda_{j} t} \int_{0}^{t} e^{\lambda_{j} s} P_{i, j-1}(s) d s, \quad j \geq i+1 .
\end{aligned}
$$

## Pure birth processes (cont.)

We start out by noting that since $P_{i j}^{\prime}(t)=-\lambda_{i} P_{i i}(t)$, it follows that

$$
\frac{P_{i i}^{\prime}(t)}{P_{i i}(t)}=\left[\ln \left(P_{i i}(t)\right)\right]^{\prime}=-\lambda_{i}
$$

By integrating both sides, we get:

$$
\ln \left(P_{i j}(t)\right)=-\lambda_{i} t+c
$$

This implies that:

$$
P_{i i}(t)=e^{-\lambda_{i} t+c}
$$

Since $P_{i i}(0)=1$, it follows that $c=0$. Hence, we get:

$$
P_{i i}(t)=e^{-\lambda_{i} t}
$$

## Pure birth processes (cont.)

In order to prove the corresponding result for $P_{i j}(t)$ where $j \geq i+1$, we assume that we have determined $P_{i, j-1}(t)$ already, and rewrite the differential equation as:

$$
P_{i j}^{\prime}(t)+\lambda_{j} P_{i j}(t)=\lambda_{j-1} P_{i, j-1}(t)
$$

We then multiply both sides by the integrating factor $e^{\lambda_{j} t}$ and get:

$$
P_{i j}^{\prime}(t) e^{\lambda_{j} t}+\lambda_{j} e^{\lambda_{j} t} P_{i j}(t)=\lambda_{j-1} e^{\lambda_{j} t} P_{i, j-1}(t) .
$$

## Pure birth processes (cont.)

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$$

Using the product rule for derivatives, the left-hand side can be simplified to:

$$
\left(P_{i j}(t) e^{\lambda_{j} t}\right)^{\prime}=\lambda_{j-1} e^{\lambda_{j} t} P_{i, j-1}(t)
$$

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$$
\left(P_{i j}(t) e^{\lambda_{j} t}\right)^{\prime}=\lambda_{j-1} e^{\lambda_{j} t} P_{i, j-1}(t)
$$

Integrating both sides of this equation we get:

$$
P_{i j}(t) e^{\lambda_{j} t}=\lambda_{j-1} \int_{0}^{t} e^{\lambda_{j} s} P_{i, j-1}(s) d s+C
$$

## Pure birth processes (cont.)

We note that:

$$
P_{i j}(t) e^{\lambda_{j} t}=\lambda_{j-1} \int_{0}^{t} e^{\lambda_{j} s} P_{i, j-1}(s) d s+C
$$

is equivalent to:

$$
P_{i j}(t)=\lambda_{j-1} e^{-\lambda_{j} t} \int_{0}^{t} e^{\lambda_{j} s} P_{i, j-1}(s) d s+C e^{-\lambda_{j} t}
$$

In order to determine the constant $C$ we use that when $j \geq i+1$, we have $P_{i j}(0)=0$. Hence, we must have $C=0$, and we conclude that:

$$
P_{i j}(t)=\lambda_{j-1} e^{-\lambda_{j} t} \int_{0}^{t} e^{\lambda_{j} s} P_{i, j-1}(s) d s
$$

## Example 6.12 - Birth and death processes

We recall that for a birth and death process with birth rates $\lambda_{0}, \lambda_{1}, \ldots$ and death rates $\mu_{1}, \mu_{2}, \ldots$ we have:

$$
\begin{aligned}
& q_{i, i+1}=\lambda_{i}, \quad i=0,1, \ldots \\
& q_{i, i-1}=\mu_{i}, \quad i=1,2, \ldots \\
& q_{i, j}=0, \quad \text { otherwise }
\end{aligned}
$$

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q_{i, i-1} & =\mu_{i}, \quad i=1,2, \ldots \\
q_{i, j} & =0, \quad \text { otherwise }
\end{aligned}
$$

Hence, we also have:

$$
\begin{aligned}
v_{0} & =\sum_{j=0}^{\infty} q_{0 j}=\lambda_{0} \\
v_{i} & =\sum_{j=0}^{\infty} q_{i j}=\lambda_{i}+\mu_{i}, \quad i=1,2, \ldots
\end{aligned}
$$

## Example 6.12 (cont.)

We insert this into the forward equations, handling the case where $j=0$ separately:

$$
\begin{aligned}
P_{i 0}^{\prime}(t) & =\sum_{k \in \mathcal{X} \backslash 0} P_{i k}(t) q_{k 0}-P_{i 0}(t) v_{0} \\
& =P_{i 1}(t) q_{10}-P_{i 0}(t) v_{0}=\mu_{1} P_{i 1}(t)-\lambda_{0} P_{i 0}(t)
\end{aligned}
$$

## Example 6.12 (cont.)

We insert this into the forward equations, handling the case where $j=0$ separately:

$$
\begin{aligned}
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& =P_{i 1}(t) q_{10}-P_{i 0}(t) v_{0}=\mu_{1} P_{i 1}(t)-\lambda_{0} P_{i 0}(t) \\
P_{i j}^{\prime}(t) & =\sum_{k \in \mathcal{X} \backslash j} P_{i k}(t) q_{k j}-P_{i j}(t) v_{j} \\
& =P_{i, j-1}(t) q_{j-1, j}+P_{i, j+1}(t) q_{j+1, j}-P_{i j}(t) v_{j} \\
& =\lambda_{j-1} P_{i, j-1}(t)+\mu_{j+1} P_{i, j+1}(t)-\left(\lambda_{j}+\mu_{j}\right) P_{i j}(t)
\end{aligned}
$$


[^0]:    ${ }^{1}$ In the Ross(2019) state 0 is the functioning state and state 1 is the failed state.

