

TD5 : Kolmogorov's Equations

Exercise 1 — Birth and death processes.

We consider the continuous time Markov process X with values in \mathbb{N} and intensity matrix Q given by:

$$q_{i,j} = \begin{cases} \beta_i & \text{si } j = i + 1 \\ \delta_i & \text{si } j = i - 1 \\ -\beta_i - \delta_i & \text{si } j = i \neq 0 \\ -\beta_i & \text{si } j = i = 0 \\ 0 & \text{sinon,} \end{cases}$$

where the β_i et δ_i are assumed to be nonnegative.

- (1) Let $i \in \mathbb{N}$, using Kolmogorov's equation, write down a system of differential equations satisfied by $p_{ij}(t) = \mathbb{P}_i(X_t = j)$.
- (2) Let λ be a probability measure on \mathbb{N} , show that when for all $i \in \mathbb{N}$, $\delta_i = 0$ the system of differential equations of the previous question subject to the initial condition $p_j(0) = \lambda(\{j\})$ admits at most one solution on \mathbb{R}_+ .
- (3) Assume that for all $i \in I$, $\beta_i = \beta$ and $\delta_i = 0$, show that when the process is started at $X_0 = 0$, the law of X_t is Poisson with parameter βt .
- (4) Assume that for all $i \in I$, $\beta_i = 0$ and $\delta_i = i\delta$, show that when the process is started at $X_0 = C > 0$, the law of X_t is binomial with parameters $(C, e^{-\delta t})$.
- (5) Give an interpretation of the Markov chain in (4), give another way to compute the value of the extinction probability $\mathbb{P}_C(X_t = 0)$.

Exercise 2 — Kolmogorov's equations makes your life easier.

Let I be a set and X a continuous time Markov chain with intensity matrix Q . Let λ be a signed measure on I and $f : I \rightarrow \mathbb{R}$, we let $g_\lambda(t) = \mathbb{E}_\lambda[f(X_t)] := \sum \lambda(\{i\}) \mathbb{E}_i[f(X_t)]$. In this exercise, we assume that the integrals/sums are well-defined and that we can derive under the integral/sum. Note this is in particular always the case when I is finite, but could be false in general.

- (1) Identifying the measure λ with the lign vector $(\lambda(\{i\}))_{i \in I}$ and the function f with the column vector $(f(j))_{j \in I}$, show we have

$$g_\lambda(t) = \lambda P(t) f.$$

- (2) Show that g_λ is differentiable and that for every $t \geq 0$,

$$g'_\lambda(t) = \mathbb{E}_{\lambda Q}[f(X_t)] = \mathbb{E}_\lambda[Q f(X_t)].$$

Consider a population of independent bacterias. Each of the bacteria splits into two bacterias after an exponential time of parameter λ . Let X_t denote the number of bacterias in the population at time t . The process X is a Markov chain with intensity matrix,

$$q_{i,j} = \begin{cases} -\lambda i & \text{when } j = i \\ \lambda i & \text{when } j = i + 1 \\ 0 & \text{otherwise} \end{cases}$$

- (3) Let $r \in (-1, 1)$, use the previous questions to find a differential equation satisfied by $g_1(t) = \mathbb{E}_1[r^{X_t}]$.
- (4) Compute the law of X_t .

Exercise 3 — *Intensity matrix and transition matrices.*

Let I be a finite set, we set that a matrix P on I is stochastic when all of its entries are nonnegative and for every $i \in I$,

$$\sum_{j \in I} P_{i,j} = 1.$$

Let $Q = (q_{i,j})_{i,j \in I}$ be a matrix on I , for $t \geq 0$, let $P(t) = e^{tQ}$. We aim to show the equivalence of the following three statements:

- (i) Q is an intensity matrix
 - (ii) $P(t)$ is a stochastic matrix for all t in a neighbourhood of 0.
 - (iii) $P(t)$ is a stochastic matrix for all $t \geq 0$.
- (1) Show (ii) and (iii) are equivalent.
- (2) Show (ii) implies (i).
- (3) We now suppose (i) is satisfied.
- (a) Show that for all i and all t , we have $\sum_j P(t)_{i,j} = 1$. (*Hint: Use the ODE satisfied by P*)
 - (b) Show the entries of the matrix $P(t)$ are nonnegative.