

Time Homogeneity

E.g. Poisson process N_t

$$q_{s,t}(n, B) = \mathbb{E}[\mathbb{I}_B(n + N_t - N_s)]$$

$$= \sum_{m \in B} \text{Poisson}(\lambda(t-s)) \{m-n\}$$

$\underbrace{\hspace{10em}}_{N(0, t-s)} \quad q_{s,t} = q_{0,t-s}$

E.g. (pre-) Brownian motion B_t

$$q_{s,t}(x, B) = \mathbb{E}[\mathbb{I}_B(x + B_t - B_s)]$$

These transition kernels $q_{s,t}$ depend on s, t only through $t-s$.

Def. A collection $\{Q_{s,t}\}_{s \leq t \in T}$ of Markov transition operators is called **time homogeneous** if

$$Q_{s,t} = Q_{0,t-s} =: Q_{t-s}, \quad \forall s \leq t \in T.$$

In this case, the Chapman-Kolmogorov equations become

$$Q_s Q_t = Q_{s+t}, \quad Q_0 = \text{Id}.$$

$$Q_{r,s} Q_{s,t} = Q_{s-r} Q_{t-s} = Q_{s-r+t-s} = Q_{t-r} = Q_{r,t}.$$

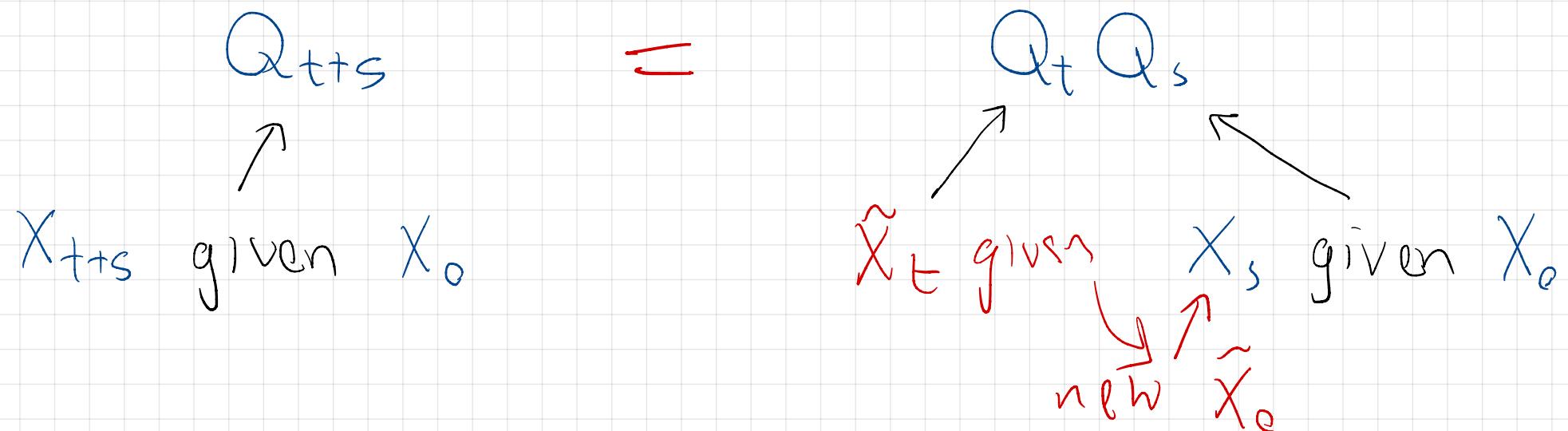
A collection of operators $(Q_t)_{t \in T}$ for which $Q_s Q_t = Q_{s+t}$, $Q_0 = \text{Id}$,
is called a **1-parameter semigroup**.

If $(X_t)_{t \in T}$ is a Markov process with time homogeneous transition operators,
it is called a **time homogeneous Markov process**.

$$\begin{aligned} & \downarrow \\ \mathbb{E}[f(X_t) | \mathcal{F}_s] &= (Q_{t-s} f)(X_s) \\ \mathbb{E}[f(X_{s+t}) | \mathcal{F}_s] &= (Q_t f)(X_s) \end{aligned}$$

Heuristic:

Q_t determines the process at time t (given X_0).



For any time homogeneous Markov process, all f.d. distributions are determined by $\nu_0 = \text{Law}(X_0)$ and the transition semigroup $(Q_t)_{t \in T}$.

Thinking of the process as a measure on path space, we fix the transition semigroup and consider the family \mathbb{P}^ν of processes with different starting distributions ν .

Notation: We let $X_\cdot = (X_t)_{t \in T}$ denote a whole family of Markov processes with given transition semigroup $(Q_t)_{t \in T}$. For $F \in \mathcal{B}(S^T, \mathcal{B}^{\otimes T})$

$E^\nu[F(X_\cdot)]$ = expected value of F (the process with $X_0 \stackrel{d}{=} \nu$) .

In the case $\nu = \delta_x$, we write E^x , P^x .

Theorem: If X_\cdot is a time homogeneous Markov process

for $F \in \mathcal{B}(S^T, \mathcal{B}^{\otimes T})$, $x \mapsto E^x[F(X_\cdot)]$ is measurable.

Moreover, for $t \geq 0$, $\nu_0 \in \text{Prob}(S, \mathcal{B})$, $E^{\nu_0}[F(X_\cdot)] = g(x)$

$$E^{\nu_0}[F(X_{t+ \cdot}) | \mathcal{F}_t] = E^{\nu_0}[F(X_{t+ \cdot}) | X_t] = \underbrace{E^{X_t}[F(X_\cdot)]}_{g(X_t)}$$

Theorem: If $X.$ is a time homogeneous Markov process, $F \in \mathcal{B}(S^{\otimes T})$

1. $S \ni x \mapsto E^x[F(X.)]$ is $\mathcal{B}/\mathcal{B}(\mathbb{R})$ -measurable \leftarrow Does not require time homog.
2. $E^{\nu_0}[F(X_{t+ \cdot}) | \mathcal{F}_t] \stackrel{\checkmark}{=} E^{\nu_0}[F(X_{t+ \cdot}) | X_t] = E^{X_t}[F(X)]$

Pf. $F(X_{t+ \cdot}) \in \mathcal{B}(\Omega, \mathcal{F}_{\geq t}^X)$. We showed in [Lecture 36.1] that if $Y \in$ then $E[Y | \mathcal{F}_t] = E[Y | X_t]$. $Y = F(X_{t+ \cdot})$

For the remainder of the proof, we take F of the form

$$F(w) = f_1(w(t_0))f_2(w(t_1)) \cdots f_n(w(t_n)) \quad 0 = t_0 < t_1 < \cdots < t_n, \quad f_j \in \mathcal{B}(S, \mathcal{B}).$$

Prove 1,2 for such F ; then extend by Dynkin.

$$\begin{aligned} 1. \quad & E^x[F(X.)] = E^x[f_0(X_{t_0}) \cdots f_n(X_{t_n})] \\ & = \int \delta_x(dx) \int q_{t_0, t_1}(x_0, dx_1) \cdots \int q_{t_{n-1}, t_n}(x_{n-1}, dx_n) f_0(x_0) \cdots f_n(x_n) \\ & = E_{\delta_x} [f_0 Q_{t_0, t_1} (\cdots f_{n-1} (Q_{t_{n-1}, t_n} f_n)^{-1})^{-1}]. \\ & = f_0(x) Q_{t_0, t_1} (\cdots f_{n-1} (Q_{t_{n-1}, t_n} f_n)^{-1})(x) \text{ meas. } \checkmark \end{aligned}$$

2. Want to show $E^{\nu_0}[F(X_{t+ \cdot})|X_t] = E^{X_t}[F(X)] = g(X_t)$
 if $g(x) = E^x[F(X \cdot)]$

$$F(X_{t+\cdot}) = f_0(X_t) f_1(X_{t+t_1}) \cdots f_n(X_{t+t_n}).$$

$\therefore \text{The } B(S, B) ,$

$$\frac{1}{V}(X_0)$$

$$E^{\nu_0}[h(X_t)F(X_{t+ \cdot})] = E^{\nu_0}[h(X_t)f_0(X_t)f_1(X_{t+t_1}) \cdots f_n(X_{t+t_n})]$$

$$= \int v_o(dx_{-1}) \int q_{o,t}(x_{-1}, dx_e) q_{t,b+b_1}(x_0, dx_1) \cdots q_{t+t_{n-1}, t+t_n}(x_{n-1}, dx_n) h(x_e) f_o(x_0) \sim f_n(x_n)$$

$$\text{Now } V_o(dx) \int q_t(x, dy) = V_t(dx) \quad \text{Law}(X_t)$$

$$Y = \int \mathcal{V}_t(dx_0) \int q_{t_1}(x_0, dx_1) \cdots q_{t_n-t_{n-1}}(x_{m-1}, dx_n) h(x_0) f_n(x_n) - f_n(x_m)$$

$$= \int h(\tau_0) V_f(dx_0) \left\{ q_{t_1}(\tau_0, dx_1) - q_{t_n-t_{n-1}}(\tau_{n-1}, dx_n) \right\} f_0(x_0) - f_n(\tau_n)$$

$$\mathbb{E}^{x_0} \left[f_n(x_0) f_n(x_t) - f_n(x_{t_n}) \right]$$

$$= \mathbb{E}^{V_0} [h(X_t) \mathbb{E}^{X_t} [F(X_+)]], \leftarrow \mathbb{E}^{x_0} [F(X_+)]$$