

# Applied Probability

ARTHUR CONMY\*

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These notes are my best attempt at making a course with ‘applied’ in the title i) exciting and ii) intuitive. Credit due to Evan Chen for the style file for these notes<sup>1</sup>.

Proof that  $\lambda > p\mu_2$  implies the number of customers is not positive recurrent:  
Consider the following independent Poisson clocks:

- $A$ , a clock with rate  $\lambda$ .
- $D_i$  (where  $i \in \mathbb{N}$ ), a clock with rate  $i\mu_2p(1-p)^{i-1}$ .

These together form a CTMC  $(X_t)$  on  $\mathbb{Z}$ , where initially the state is  $+1$ , when an  $A$  clock rings we transition  $+1$ , and when a  $D_i$  clock rings we transition  $-1$ . We now want to perform two steps:

1. Couple  $(X_t)$  with  $(M_t + N_t)$  (which will be ‘greater than’ it).
2. Show  $(X_t)$  is not positive recurrent.

For 1., consider  $X_0 = 1$ , and  $M_0 = 0, N_0 = 1$ .

**Exercise 0.1.** Find a process  $(X_t)_{t \geq 0}$  that has stationary but not independent increments, and a process that has independent but not stationary increments.

- Stationary but not independent: take some  $\text{Poi}(1)$  random variable, and when it fires, jump some random  $U[0, 1]$  distance.
- Independent but not stationary: deterministic thing which doesn’t have fixed jumps e.g  $0, 1, 0, 1, \dots$  on intervals of length 1.

## Theorem 0.2

The following are equivalent:

- $(X_t)$  is a CTMC with generator  $Q$ .
- $(X_t)$  satisfies the limiting transition property that
  - When  $y \neq x$ ,  $\mathbb{P}(X_{t+h} = y | X_t = x) = hq_{xy} + O(h^2)$ .
  - $\mathbb{P}(X_{t+h} = x | X_t = x) = 1 - h \sum_{y \neq x} q_{xy} + O(h^2)$ .

\*Please send any corrections and/or feedback to [asc70@cam.ac.uk](mailto:asc70@cam.ac.uk)

<sup>1</sup>Available here: <https://github.com/vEnhance/dotfiles/blob/master/texmf/tex/latex/evan/evan.sty>.

## §1 Reversibility

What happens when we run CTMCs in reverse? We will restrict to the case  $(Q, \pi)$  where we have an invariant distribution, much like we did in IB Markov Chains.

### Theorem 1.1 (Reversibility)

Let the irreducible, non-explosive CTMC  $(X_t)$  with generator  $Q$  have invariant distribution  $\pi$ . Then fixing some constant end time  $T > 0$ , the random process  $(\hat{X}_t) = (X_{T-t})$  for  $0 \leq t \leq T$  is a CTMC, with invariant distribution

$$\hat{q}_{xy} = \frac{\pi_y}{\pi_x} q_{yx}. \quad (1)$$

**Remark 1.2.** This is intuitively the case, since the distribution of  $X_T$  will also follow the invariant distribution, and  $\pi_x \hat{q}_{xy} = \pi_y q_{yx}$  encodes the fact that the reversed process will have transitions from state  $x$  to state  $y$  identical to the transitions from state  $y$  to state  $x$  in the original process.

We can bash this claim out with a lot of algebra, although building from the remark, we can see the result as a case of Bayes' theorem.

*Proof.* Since  $\pi$  is invariant,  $\forall t \in [0, T]$ ,  $\hat{X}_t$  has distribution  $\pi$ . Now fix such a  $t$  - we go for the second characterisation of (0.2).

For  $h$  small, and  $y \neq x$ , by Bayes' theorem,

$$\mathbb{P}(\hat{X}_{t+h} = y | \hat{X}_t = x) = \mathbb{P}(X_{T-t-h} = y | X_{T-t} = x) = \frac{\mathbb{P}(X_{T-t} = x | X_{T-t-h} = y) \mathbb{P}(X_{T-t-h} = y)}{\mathbb{P}(X_{T-t} = x)} \quad (2)$$

which equals  $\frac{h\pi_y q_{yx}}{\pi_x}$ . Therefore indeed  $\hat{X}$  is a CTMC. □