

Mean and Variance Calculation for Finite State Birth and Death Processes

We give two related approaches to finding the mean and variance of a continuous time birth and death process with finite state space. The basic set up is that the state space S is $S = \{0, 1, 2, \dots, N\}$ for some $N \geq 1$, that

$$\begin{aligned}\Pr(X_{t+h} = s + 1 | X_t = s) &= \lambda(s)h + o(h) \\ \Pr(X_{t+h} = s - 1 | X_t = s) &= \delta(s)h + o(h) \\ \Pr(X_{t+h} = s | X_t = s) &= 1 - (\lambda(s) + \delta(s))h + o(h)\end{aligned}$$

We assume that $\delta(s) > 0$ if $s \neq 0$ and $\delta(0) = 0$, while $\lambda(s) > 0$ if $s \neq N$ and $\lambda(N) = 0$. We are particularly interested in the case where $\lambda(\cdot)$ and $\delta(\cdot)$ are polynomials of degree at most two, for then we can outline a direct method for computing $E[X_t]$ and $E[X_t^2]$ without computing $\Pr(X_t = s)$.

Remember that

$$\begin{aligned}E[X_t] &= \sum_{s \in S} s \Pr(X_t = s) \\ E[X_t^2] &= \sum_{s \in S} s^2 \Pr(X_t = s)\end{aligned}$$

Since $P_t = \exp(t\Lambda)$ we know that both $E[X_t]$ and $E[X_t^2]$ are differentiable functions of t .

1 Differential Equation Method

We will construct differential equations for the mean and second moment of X_t .

1.1 First Moment

Suppose that $\lambda(s) - \delta(s) = a - bs$. Since $\lambda(N) = \delta(0) = 0$ we have

$$\begin{aligned}\lambda(0) &= a \\ -\delta(N) &= a - bN\end{aligned}$$

so

$$\begin{aligned}a &= \lambda(0) \\ b &= \frac{\lambda(0) + \delta(N)}{N}.\end{aligned}$$

In particular, both a and b are positive.

Since there are only a finite number of states, all the $o(h)$ terms will always recombine to a single $o(h)$ term.

Therefore,

$$\begin{aligned}E[X_{t+h} | X_t = s] &= \sum_{k=-1}^1 (s+k) \Pr(X_{t+h} = s+k | X_t = s) + o(h) \\ &= s - \Pr(X_{t+h} = s-1 | X_t = s) + \Pr(X_{t+h} = s+1 | X_t = s) + o(h) \\ &= s + (\lambda(s) - \delta(s))h + o(h) \\ &= s + (a - bs)h + o(h)\end{aligned}\tag{1}$$

Put $M_1(t) = E[X_t]$. It follows from (1) that

$$\begin{aligned}
M_1(t+h) &= \sum_{s \in S} E[X_{t+h}|X_t = s] \Pr(X_t = s) \\
&= \sum_{s \in S} (s + (a - bs)h) \Pr(X_t = s) + o(h) \\
&= \sum_{s \in S} s \Pr(X_t = s) + h \sum_{s \in S} (a - bs) \Pr(X_t = s) + o(h) \\
&= M_1(t) + ah - bhM_1(t) + o(h)
\end{aligned} \tag{2}$$

Let $h \rightarrow 0^+$ and (since M_1 is differentiable) we get

$$M_1'(t) + bM_1(t) = a$$

which has as its solution

$$M_1(t) = \left(M_1(0) - \frac{a}{b} \right) \exp(-bt) + \frac{a}{b} \tag{3}$$

In particular, as $t \rightarrow \infty$ we have $M_1(t)$ converging to a/b , and

$$\frac{a}{b} = \frac{N\lambda(0)}{\lambda(0) + \delta(N)}.$$

1.2 Second Moment

Suppose then that we want to extend this method to second moments. This will require that $\lambda(s) - \delta(s)$ is a polynomial of degree 1 (or less) in s while $\lambda(s) + \delta(s)$ is a polynomial of degree 2 (or less) in s . This forces both $\lambda(s)$ and $\delta(s)$ to be polynomials of degree 2 (or less) in s , and, if both are quadratic, they have to have the same leading coefficient. Hence we must have

$$\begin{aligned}
\delta(s) &= As + Cs^2 \\
\lambda(s) &= B(N - s) + C(N - s)^2
\end{aligned}$$

This means that there are constants a, b, c, d expressible in terms of A, B, C, N so that

$$\begin{aligned}
\lambda(s) - \delta(s) &= a - bs \\
\lambda(s) + \delta(s) &= a - ds + cs^2
\end{aligned}$$

We now calculate in a manner parallel to what we did for $M_1(t)$.

$$\begin{aligned}
E[X_{t+h}^2|X_t = s] &= \sum_{k=-1}^1 (s+k)^2 \Pr(X_{t+h} = s+k|X_t = s) + o(h) \\
&= \sum_{k=-1}^1 (s^2 + 2ks + k^2)^2 \Pr(X_{t+h} = s+k|X_t = s) + o(h) \\
&= s^2 - 2s(\lambda(s) - \delta(s))h + (\lambda(s) + \delta(s))h + o(h) \\
&= s^2 - 2s(a - bs)h + (a - ds + cs^2)h + o(h) \\
&= s^2 + (2b + c)s^2h - (d + 2a)sh + ah + o(h)
\end{aligned}$$

Therefore

$$\begin{aligned}
M_2(t+h) &= \sum_{s \in S} \mathbb{E}[X_{t+h}^2 | X_t = s] \Pr(X_t = s) \\
&= \sum_{s \in S} (s^2 + (2b+c)s^2h - (2a+d)sh + ah + o(h)) \Pr(X_t = s) \\
&= M_2(t) + (2b+c)M_2(t)h - (2a+d)M_1(t)h + ah + o(h)
\end{aligned} \tag{4}$$

Therefore

$$\frac{M_2(t+h) - M_2(t)}{h} = (2b+c)M_2(t) - (2a+d)M_1(t) + a + \frac{o(h)}{h}.$$

Letting $h \rightarrow 0^+$ we see that

$$M_2'(t) = (2b+c)M_2(t) - (2a+d)M_1(t) + a. \tag{5}$$

Since $M_1(t)$ is known we may solve this differential equation as well by integrating factors.

1.3 Parameter restrictions

We know from the discussion in the previous section that we must have

$$\begin{aligned}
2\lambda(s) &= 2a + (b+d)s + cs^2 \\
2\delta(s) &= (d-b)s + cs^2.
\end{aligned}$$

We also must have

$$\begin{aligned}
2\lambda(s) &> 0 \text{ if } s < N \\
2\lambda(N) &= 0 \\
2\delta(s) &> 0 \text{ if } s > 0
\end{aligned}$$

so it is convenient to write

$$\begin{aligned}
\lambda(s) &= A(N-s) + C(N-s)^2 \\
\delta(s) &= Bs + Cs^2.
\end{aligned}$$

We can, by comparing coefficients, translate from a, b, c, d to A, B, C . First, we will explore the restrictions on A, B, C .