

Question: Describe explicitly the limiting procedure to obtain the diffusion process of Example 13.3.12 ("Diffusion approximation to the branching process."). Find hitting probabilities for the resulting diffusion, namely, the probability that the diffusion process hits a before hitting b if begun at x , for each a, b , and x .

Solution: Let $(Z(t))_{t \geq 0}$ be the continuous time version of the branching process. That is, a branching process in which each individual gives birth with rate 1. Let μ be the mean number of offspring per birth event and let σ^2 be the variance of the number of offspring per birth event.

To find the appropriate scaling for the diffusion limit, set $X(t) = \frac{1}{N^\beta} Z(N^\gamma t)$. Since we will be sending N to infinity, all that matters is the ratio of β to γ , so without loss of generality we may set $\gamma = 1$, so that

$$X(t) = \frac{1}{N^\beta} Z(Nt).$$

We first find the infinitesimal mean,

$$\mu_X(x, t) = \lim_{h \downarrow 0} \frac{1}{h} \mathbb{E}(X(t+h) - X(t) | X(t) = x)$$

Recall that if there are m individuals alive in a branching process then the next birth event happens with rate m , so given $X(t) = x$, i.e. given $Z(Nt) = xN^\beta$, the birth rate before time has been scaled would be xN^β , but the scaling of time by N means that births happen at rate $xN^{1+\beta}$. Using the fact that the expected change in the original population given one birth event has occurred is $\mu - 1$, so that the change in the scaled population is $\frac{\mu-1}{N^\beta}$, we have:

$$\begin{aligned} \mu_X(x, t) &= \lim_{h \downarrow 0} \frac{1}{h} \mathbb{E}(X(t+h) - X(t) | X(t) = x) \\ &= \lim_{h \downarrow 0} \frac{1}{h} \mathbb{E}(N^{-\beta} Z(Nt + Nh) - N^{-\beta} Z(Nt) | N^{-\beta} Z(Nt) = x) \\ &= xN^{1+\beta} \left(\frac{\mu-1}{N^\beta} \right) \\ &= xN(\mu - 1) \end{aligned}$$

To get a non-trivial limit for large N (i.e. not zero or infinity), we must have $\mu - 1 = O(N^{-1})$. Set $\mu = 1 + \alpha N^{-1}$, so that $\mu_X(x, t) = \alpha x$.

We now find the infinitesimal variance:

$$\begin{aligned} \sigma_X^2(x, t) &= \lim_{h \downarrow 0} \frac{1}{h} \mathbb{E}((X(t+h) - X(t))^2 | X(t) = x) \\ &= \lim_{h \downarrow 0} \frac{1}{h} \mathbb{E}((N^{-\beta} Z(Nt + Nh) - N^{-\beta} Z(Nt))^2 | N^{-\beta} Z(Nt) = x) \\ &= xN^{1+\beta} \left(\frac{\sigma^2 + (\mu-1)^2}{N^{2\beta}} \right) \\ &= xN^{1-\beta} (\sigma^2 + (\mu - 1)^2) \\ &= xN^{1-\beta} (\sigma^2 + \alpha^2 N^{-2}) \\ &\approx xN^{1-\beta} \sigma^2 \end{aligned}$$

To get a non-trivial limit for large N (i.e. not zero or infinity), whilst keeping σ^2 greater than zero and finite, we must have $1 - \beta = 0$, i.e. $\beta = 1$. Then $\sigma_X^2(x, t) = x\sigma^2$.

The calculations above can be expanded so that it is clear what is happening in the limit as $h \rightarrow 0$:

$$\begin{aligned}
& \mu_X(x, t) \\
&= \lim_{h \downarrow 0} \frac{1}{h} \mathbb{E}(X(t+h) - X(t) | X(t) = x) \\
&= \lim_{h \downarrow 0} \frac{1}{h} \mathbb{E}(N^{-\beta} Z(Nt + Nh) - N^{-\beta} Z(Nt) | N^{-\beta} Z(Nt) = x) \\
&= \lim_{h \downarrow 0} \frac{1}{h} \mathbb{P}(S \text{ jumps once during } (Nt, Nt + Nh) | Z(Nt) = xN^\beta) \\
&\quad \times \mathbb{E}(N^{-\beta} Z(Nt + Nh) - N^{-\beta} Z(Nt) | N^{-\beta} Z(Nt) = x, Z \text{ jumps once during } (Nt, Nt + Nh)) \\
&\quad + \lim_{h \downarrow 0} \frac{1}{h} \mathbb{P}(Z \text{ jumps } > \text{ once during } (Nt, Nt + Nh) | Z(Nt) = xN^\beta) \\
&\quad \times \mathbb{E}(N^{-\beta} Z(Nt + Nh) - N^{-\beta} Z(Nt) | N^{-\beta} Z(Nt) = x, Z \text{ jumps } > \text{ once during } (Nt, Nt + Nh)) \\
&= \lim_{h \downarrow 0} \frac{1}{h} (xN^\beta N h e^{-xN^\beta N h}) \\
&\quad \times \mathbb{E}(N^{-\beta} Z(Nt + Nh) - N^{-\beta} Z(Nt) | N^{-\beta} Z(Nt) = x, Z \text{ jumps once during } (Nt, Nt + Nh)) \\
&\quad + \lim_{h \downarrow 0} \frac{1}{h} O(h^2) \\
&\quad \times \mathbb{E}(N^{-\beta} Z(Nt + Nh) - N^{-\beta} Z(Nt) | N^{-\beta} Z(Nt) = x, Z \text{ jumps } > \text{ once during } (Nt, Nt + Nh)) \\
&= \lim_{h \downarrow 0} xN^{\beta+1} \mathbb{E}(N^{-\beta} Z(Nt + Nh) - N^{-\beta} Z(Nt) | N^{-\beta} Z(Nt) = x, Z \text{ jumps once during } (Nt, Nt + Nh)) \\
&\quad + 0 \\
&= xN \mathbb{E}(Z(Nt + Nh) - Z(Nt) | N^{-\beta} Z(Nt) = x, Z \text{ jumps once during } (Nt, Nt + Nh)) \\
&= xN(\mu - 1)
\end{aligned}$$

And the expanded calculations for the infinitesimal variance are:

$$\begin{aligned}
& \sigma_X^2(x, t) \\
&= \lim_{h \downarrow 0} \frac{1}{h} \mathbb{E}((X(t+h) - X(t))^2 | X(t) = x) \\
&= \lim_{h \downarrow 0} \frac{1}{h} \mathbb{E}((N^{-\beta} Z(Nt+Nh) - N^{-\beta} Z(Nt))^2 | N^{-\beta} Z(Nt) = x) \\
&= \lim_{h \downarrow 0} \frac{1}{h} \mathbb{P}(S \text{ jumps once during } (Nt, Nt+Nh) | Z(Nt) = xN^\beta) \\
&\quad \times \mathbb{E}((N^{-\beta} Z(Nt+Nh) - N^{-\beta} Z(Nt))^2 | N^{-\beta} Z(Nt) = x, Z \text{ jumps once during } (Nt, Nt+Nh)) \\
&\quad + \lim_{h \downarrow 0} \frac{1}{h} \mathbb{P}(Z \text{ jumps } > \text{ once during } (Nt, Nt+Nh) | Z(Nt) = xN^\beta) \\
&\quad \times \mathbb{E}((N^{-\beta} Z(Nt+Nh) - N^{-\beta} Z(Nt))^2 | N^{-\beta} Z(Nt) = x, Z \text{ jumps } > \text{ once during } (Nt, Nt+Nh)) \\
&= \lim_{h \downarrow 0} \frac{1}{h} (xN^\beta N h e^{-xN^\beta N h}) \\
&\quad \times \mathbb{E}((N^{-\beta} Z(Nt+Nh) - N^{-\beta} Z(Nt))^2 | N^{-\beta} Z(Nt) = x, Z \text{ jumps once during } (Nt, Nt+Nh)) \\
&\quad + \lim_{h \downarrow 0} \frac{1}{h} O(h^2) \\
&\quad \times \mathbb{E}((N^{-\beta} Z(Nt+Nh) - N^{-\beta} Z(Nt))^2 | N^{-\beta} Z(Nt) = x, Z \text{ jumps } > \text{ once during } (Nt, Nt+Nh)) \\
&= \lim_{h \downarrow 0} xN^{\beta+1} \mathbb{E}((N^{-\beta} Z(Nt+Nh) - N^{-\beta} Z(Nt))^2 | N^{-\beta} Z(Nt) = x, Z \text{ jumps once during } (Nt, Nt+Nh)) \\
&\quad + 0 \\
&= xN^{1-\beta} \mathbb{E}((Z(Nt+Nh) - Z(Nt))^2 | N^{-\beta} Z(Nt) = x, Z \text{ jumps once during } (Nt, Nt+Nh)) \\
&= xN^{1-\beta} (\sigma^2 + (\mu - 1)^2) \\
&= xN^{1-\beta} (\sigma^2 + \alpha^2 N^{-2}) \\
&\approx xN^{1-\beta} \sigma^2
\end{aligned}$$

To find hitting probabilities, we look for martingales associated with our diffusion X . X has generator

$$Gf(x) = \alpha x f'(x) + \frac{1}{2} \sigma^2 x f''(x)$$

If we find h such that $Gh(x) = 0$ for all x , then $h(X(t))$ is a martingale. Solutions to

$$\alpha x h'(x) + \frac{1}{2} \sigma^2 x h''(x) = 0$$

are found by cancelling the x , then integrating to get

$$\alpha h(x) + \frac{1}{2} \sigma^2 h'(x) = c$$

for some constant c . i.e. $h'(x) = \frac{2\alpha}{\sigma^2} h(x) + \frac{2c}{\sigma^2}$, which has solution

$$h(x) = A e^{\frac{2\alpha}{\sigma^2} x} + B$$

for some constants A and B . We choose to set $A = 1$ and $B = 0$ to see that

$$M(t) := e^{\frac{2\alpha}{\sigma^2} X(t)}$$

is a martingale. Suppose $X_0 = x$ with $0 < a < x < b$ and let $T = \inf\{t > 0 : X(t) \in \{a, b\}\}$. Let

$$p_a = \mathbb{P}(X_T = a) = 1 - p_b = 1 - \mathbb{P}(X_T = b)$$

Then by applying the optional stopping theorem, which applies since $\mathbb{P}(T < \infty) = 1$ and because $M(t)$ is uniformly bounded for $t < T$, we have

$$e^{\frac{2\alpha}{\sigma^2} x} = \mathbb{E}(M(0)) = \mathbb{E}(M(T)) = p_a e^{\frac{2\alpha}{\sigma^2} a} + p_b e^{\frac{2\alpha}{\sigma^2} b}$$

Using $p_a = 1 - p_b$ we can solve this to get

$$p_a = \frac{e^{\frac{2\alpha}{\sigma^2} x} - e^{\frac{2\alpha}{\sigma^2} b}}{e^{\frac{2\alpha}{\sigma^2} a} - e^{\frac{2\alpha}{\sigma^2} b}}$$

and

$$p_a = \frac{e^{\frac{2\alpha}{\sigma^2} a} - e^{\frac{2\alpha}{\sigma^2} x}}{e^{\frac{2\alpha}{\sigma^2} a} - e^{\frac{2\alpha}{\sigma^2} b}}$$