



Stochastic population modeling

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3.8.1 The Green function $G(x,x_0)$ for a diffusion process

The expected time the process is in $(x, x + \Delta x)$ before reaching an absorbing barrier is $G(x, x_0) \Delta x$ when $\Delta x \rightarrow 0$.

Expected time in $[c, d]$:
$$\int_c^d G(x, x_0) dx$$

Mathematical expressions for $G(x, x_0)$

$$s(x) = \exp \left[-2 \int_{-\infty}^x \frac{\mu(z)}{\nu(z)} dz \right] \quad \text{and} \quad S(x) = \int_{-\infty}^x s(z) dz$$

Help function (to simplify notation): $m(x) = \frac{1}{\nu(x)s(x)}$,

$$G(x, x_0) = \begin{cases} 2 \frac{[S(x) - S(a)][S(b) - S(x_0)]}{S(b) - S(a)} m(x) & a \leq x \leq x_0 \leq b \\ 2 \frac{[S(b) - S(x)][S(x_0) - S(a)]}{S(b) - S(a)} m(x) & a \leq x_0 \leq x \leq b \end{cases}$$

Exponential growth with demographic variance

$$\begin{aligned}\mu_N(n) &= rn \\ \nu_N(n) &= \sigma_d^2 n + \sigma_e^2 n^2\end{aligned} \Rightarrow s(n) = \left(\frac{\sigma_d^2 + \sigma_e^2}{\sigma_d^2 + \sigma_e^2 n} \right)^{\frac{2r}{\sigma_e^2}}$$

$$\Rightarrow S(n) = \begin{cases} \frac{\sigma_d^2 + \sigma_e^2}{2s} \left[1 - \left(\frac{\sigma_d^2 + \sigma_e^2 n}{\sigma_d^2 + \sigma_e^2} \right)^{\frac{2s}{\sigma_e^2}} \right] & s \neq 0 \\ \frac{\sigma_d^2 + \sigma_e^2}{2s} \ln \left(\frac{\sigma_d^2 + \sigma_e^2 n}{\sigma_d^2 + \sigma_e^2} \right) & s = 0 \end{cases}$$

With $b = \infty$, $P(\text{extinction}) = 1$ for $s < 0$

and

$$P(\text{ext}) = \left(\frac{\sigma_d^2 + \sigma_e^2}{\sigma_d^2 + \sigma_e^2 n_0} \right)^{\frac{2s}{\sigma_e^2}} \quad s > 0$$

Fig. 3.4

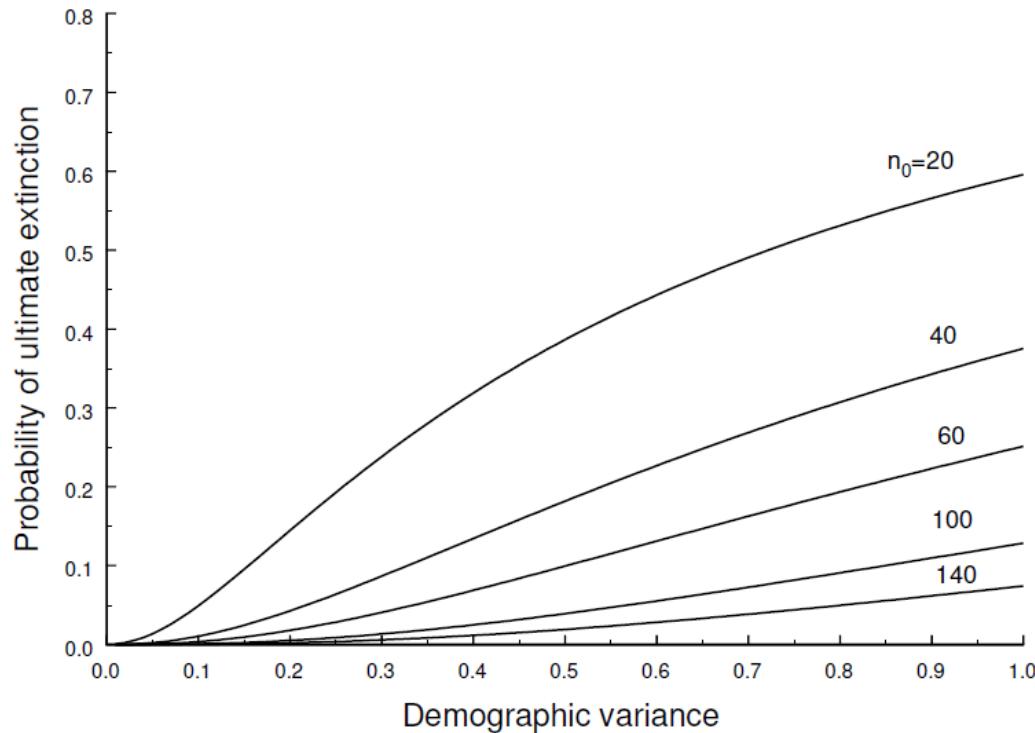


Figure 3.4: The probability of ultimate extinction as function of the demographic variance for different initial population sizes. The diffusion has infinitesimal mean $\mu(n) = rn$ with $r = 0.015$ and constant environmental and demographic variance, $\sigma_e^2 = 0.01$.

Expected time to extinction, T

For populations with density regulation, let a=1
and b infinity:

$$E(T) = \int_1^{n_0} 2m(n) S(n) dn + \int_{n_0}^{\infty} 2m(n) S(n_0) dn$$

Stationary distributions - OU

Adds density regulation (no extinction barrier):

$$\mu(x) = \alpha - \beta x$$

$$\nu(x) = \sigma^2$$

$$\Rightarrow X_t | X_0 = x_0 \sim N\left(\frac{\alpha}{\beta} + \left(x_0 - \frac{\alpha}{\beta}\right)e^{-\beta t}, \frac{\sigma^2}{2\beta}(1 - e^{-2\beta t})\right)$$

Stationary distribution ($t \rightarrow \infty$): $f_\infty(x, x_0) = f(x)$

Here: $N\left(\frac{\alpha}{\beta}, \frac{\sigma^2}{2\beta}\right)$

Stationary distributions – theta-logistic model

$$\theta \neq 0, \quad \mu(n) = rn \left(1 - \left(\frac{n}{K} \right)^\theta \right), \quad r = \frac{r_1}{(1 - K^{-\theta})}$$

$$\nu(n) = \sigma_e^2 n^2$$

$$\Rightarrow f(n; \alpha, K, \theta) = Cn^{\alpha-1} e^{-\frac{(\alpha+1)}{\theta} \left(\frac{n}{K} \right)^\theta}, \quad \alpha = \frac{2r}{\sigma_e^2} - 1 = \frac{2s}{\sigma_e^2} > 0$$

Transformation of variable: $y = \left(\frac{n}{K} \right)^\theta$

$$\Rightarrow f(n; \alpha, K, \theta) = \frac{|\theta| \left(\frac{\alpha+1}{\theta} \right)^{\alpha/\theta}}{K \Gamma(\alpha/\theta)} \left(\frac{n}{K} \right)^{\alpha-1} e^{-\frac{(\alpha+1)}{\theta} \left(\frac{n}{K} \right)^\theta}, \quad \theta \neq 0$$

This is called the generalized gamma dist.

Fig. 3.5

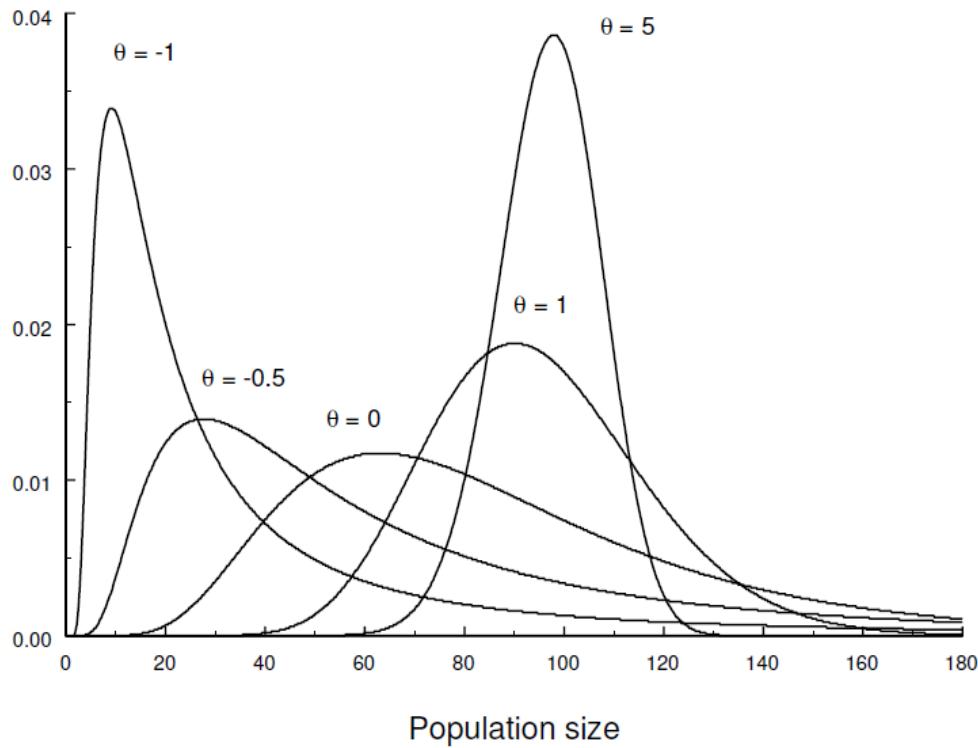


Figure 3.5: The stationary distribution for the theta-logistic model, the generalized gamma distribution, for different values of θ . The other parameters are $K = 100$, $\sigma_e^2 = 0.01$, and $r_1 = 0.1$.

