Lecture 5: Immigration-emigration models #4

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30 May 2011

1 Probability generating function

The population size n is a discrete random number and is associated with a probability distribution $P_n(t)$ which evolves with time t according to master equation. Solving $P_n(t)$ in general is not always easy but for some simple case we can do it. We now introduce **probability generating function** G(t, z) associated with a probability distribution $P_n(t)$ as

$$G(t,z) = \sum_{n} P_n(t) z^n \tag{1}$$

where the summation is taken for all possible n. This series will converge when |z| < r where r is convergence radius. Probability generating function is just a power series whose *i*-th coefficient is $P_i(t)$.

Let's take a look of general properties of probability generating function. It is obvious that G(t, z) evaluated at z = 1 is always 1 because it is just summation of probability $P_n(t)$.

$$G(t,1) = \sum_{n} P_n(t) = 1$$
 (2)

Differentiating (2) with z yields

$$\frac{\partial}{\partial z}G(t,z) = \sum_{n} nP_n(t)z^{n-1}$$

and by substituting z = 1 we have a useful result

$$\frac{\partial}{\partial z}G(t,z)\bigg|_{z=1} = \sum_{n} nP_n(t) = \langle n \rangle = E[n]$$
(3)

Similarly differentiating (2) with z twice

$$\frac{\partial^2}{\partial z^2}G(t,z) = \sum_n n(n-1)P_n(t)z^{n-2}$$

and substituting z = 1 gives

$$\frac{\partial^2}{\partial z^2} G(t,z) \bigg|_{z=1} = \sum_n n(n-1) P_n(t) = E[n(n-1)] = \langle n^2 \rangle - \langle n \rangle$$
(4)

We now remember that $Var[n] = \langle n^2 \rangle - \langle n \rangle^2$. That is,

$$Var[n] = \left\{ \frac{\partial^2}{\partial z^2} G(t,z) + \frac{\partial}{\partial z} G(t,z) - \left(\frac{\partial}{\partial z} G(t,z)\right)^2 \right\} \bigg|_{z=1}$$
(5)

These calculations show that if we can solve and obtain a probability generating function G(t, z), we can derive the ensemble average $\langle n \rangle$ and the variance Var[n] from G(t, z). In previous lectures, we derived moment dynamics directly from master equation. But they can be also obtained from G(t, z). Even more, probability $P_n(t)$ is given as a coefficient of Taylor expansion of G(t, z) around z = 0. This means solving G(t, z) is equivalent to solving $P_n(t)$. In the following sections we try to solve the p.g.f. G(z, t) of the stochastic immigration-emigration process.

2 Solving the pgf of immigration-emigration process

We now solve the pgf of immigration-emigration process in which n can be negative (population size is no longer restricted non-negative). The master equation is

$$\frac{dP_n(t)}{dt} = \alpha P_{n-1}(t) + \beta P_{n+1}(t) - (\alpha + \beta)P_n(t) \quad \text{for } -\infty < n < \infty$$
(6)

and the pgf is defined as

$$G(t,z) = \sum_{n} P_n(t) z^n \tag{7}$$

Differentiating the pgf with t yields

$$\frac{\partial}{\partial t}G(t,z) = \sum_{n} \frac{d}{dt} P_n(t) z^n$$

Using the master equation (6), we have

$$\frac{\partial}{\partial t}G(t,z) = \sum_{n} \left\{ \alpha P_{n-1}(t) + \beta P_{n+1}(t) - (\alpha + \beta)P_n(t) \right\} z^n$$
$$= \alpha z \sum_{n} P_{n-1}(t)z^{n-1} + \frac{\beta}{z} \sum_{n} P_{n+1}(t)z^{n+1} - (\alpha + \beta) \sum_{n} P_n(t)z^n$$
$$= (\alpha z + \beta/z - \alpha - \beta)G(t,z)$$

This is ODE of G(t, z) with respect to time t and has unique solution with initial condition $G(0, z) = z^m$ where m is initial population size at t = 0, n(0). This can be readily solved by variable separation. The solution is

$$G(t,z) = z^m \exp\left[(-\alpha - \beta + \alpha z + \beta/z)t\right]$$
(8)

3 Problem

We have solved the pgf of the immigration-emigration process G(t, z).

- 1. Confirm that the expected value E[n] and variance Var[n] of n derived from the moment dynamics in the last lecture coincide with those derived from the pgf.
- 2. By Taylor expanding the pgf G(t, z) with respect to z and looking at coefficients of z^n , $P_n(t)$ can be obtained. This is not actually easy but we are a bit close to the solution $P_n(t)$.