

Solving PDE of the type

$$\dot{P}(z, t) = a(z) \cdot P'(z, t)$$

where $a(z)$ is a specific (given) function of z .

First we note that, if $P(z, t)$ is a solution, then any function of $P(z, t)$, say $g[P(z, t)]$ is also a solution. This is clear from the chain rule:

$$\begin{aligned}\dot{g}[P(z, t)] &= \dot{g}[P(z, t)] \cdot \dot{P}(z, t) \\ g'[P(z, t)] &= g'[P(z, t)] \cdot P'(z, t)\end{aligned}$$

Substituted back into the original equation, $\dot{g}[P(z, t)]$ cancels out.

We will assume that

$$P(z, t) = Q(z) \cdot R(t)$$

Substitute and get:

$$Q(z) \cdot \dot{R}(t) = a(z) \cdot Q'(z) \cdot R(t)$$

Divide by $Q(z) \cdot R(t)$:

$$\frac{\dot{R}(t)}{R(t)} = a(z) \frac{Q'(z)}{Q(z)}$$

A function of t (no z) can be equal to a function of z (no t) only if both of them are equal to the same constant, say $-\gamma$. We thus get

$$\frac{\dot{R}(t)}{R(t)} = -\gamma$$

and

$$a(z) \frac{Q'(z)}{Q(z)} = -\gamma$$

The first of these has the following general solution

$$R(t) = c \cdot e^{-\gamma t}$$

the second one implies

$$\ln Q(z) = -\gamma \int \frac{dz}{a(z)}$$

or

$$Q(z) = \exp\left(-\gamma \int \frac{dz}{a(z)}\right)$$

We then know that

$$g\left[c \cdot e^{-\gamma t} \cdot \exp\left(-\gamma \int \frac{dz}{a(z)}\right)\right]$$

is also a solution, where g is any univariate function. Clearly, both the multiplication by c and raising to the power of γ can be 'absorbed' into g , so re-writing the solution as

$$g\left[e^{-t} \cdot \exp\left(-\int \frac{dz}{a(z)}\right)\right]$$

is basically saying the same thing. Furthermore, one can show that this is the general solution of the original equation (i.e. all its solutions have this form).

The initial condition

$$P(z, 0) = z^i = g \left[\exp \left(- \int \frac{dz}{a(z)} \right) \right]$$

(where i is the value of the process at time $t = 0$) then determines the exact form of $g(\cdot)$.

EXAMPLES

1. Yule process, $a(z) = -\lambda z(1 - z)$. Since

$$\int \frac{dz}{z(1-z)} = \int \left(\frac{1}{z} + \frac{1}{1-z} \right) dz = \ln z - \ln(1 - z)$$

we get

$$\begin{aligned} P(z, t) &= g_0 \left[e^{-t} \cdot \exp \left(\frac{1}{\lambda} \ln \frac{z}{1-z} \right) \right] = \\ &g_0 \left[e^{-t} \cdot \left(\frac{z}{1-z} \right)^{1/\lambda} \right] \equiv g \left[e^{-\lambda t} \cdot \frac{z}{1-z} \right] \end{aligned}$$

where $g(\cdot)$ is such that

$$g \left(\frac{z}{1-z} \right) = z^i$$

or

$$g(x) = \left(\frac{x}{1+x} \right)^i$$

The final solution is thus

$$P(z, t) = \left[\frac{e^{-\lambda t} \cdot \frac{z}{1-z}}{1 + e^{-\lambda t} \cdot \frac{z}{1-z}} \right]^i = \left[\frac{e^{-\lambda t} \cdot z}{1 - z + e^{-\lambda t} \cdot z} \right]^i$$

This is clearly a PGF of the negative binomial distribution (number of trials to get the i^{th} success) with $p_t = e^{-\lambda t}$.

2. Pure-Death process, $a(z) = \mu(1 - z)$.

$$\begin{aligned} P(z, t) &= g_0 \left\{ e^{-t} \cdot \exp \left[\frac{1}{\mu} \ln(1 - z) \right] \right\} = \\ &g_0 \left[e^{-t} \cdot (1 - z)^{1/\mu} \right] \equiv g \left[e^{-\mu t} \cdot (1 - z) \right] \end{aligned}$$

where $g(\cdot)$ is such that

$$g(1 - z) = z^i$$

or

$$g(x) = (1 - x)^i$$

The final solution is thus

$$P(z, t) = [1 - e^{-\mu t} \cdot (1 - z)]^i = (1 - e^{-\mu t} + e^{-\mu t} \cdot z)^i$$

This is clearly a PGF of the binomial distribution, with $p_t = e^{-\mu t}$ and total number of trials equal to i .

3. Linear-Growth process: $a(z) = (1 - z)(\mu - \lambda z)$. Since

$$\int \frac{dz}{(1-z)(\mu-\lambda z)} = \frac{1}{\mu-\lambda} \int \left(\frac{1}{1-z} - \frac{\lambda}{\mu-\lambda z} \right) dz = \frac{1}{\mu-\lambda} [\ln(\mu - \lambda z) - \ln(1 - z)]$$

we get

$$P(z, t) = g_0 \left[e^{-t} \cdot \left(\frac{1-z}{\mu-\lambda z} \right)^{\frac{1}{\mu-\lambda}} \right] = g \left(e^{-t(\mu-\lambda)} \cdot \frac{1-z}{\mu-\lambda z} \right)$$

where $g(\cdot)$ is such that

$$g\left(\frac{1-z}{\mu-\lambda z}\right) = z^i$$

or

$$g(x) = \left(\frac{1-\mu x}{1-\lambda x}\right)^i$$

The final solution is thus

$$\begin{aligned} P(z, t) &= \left[\frac{1 - \mu \cdot e^{-t(\mu-\lambda)} \cdot \frac{1-z}{\mu-\lambda z}}{1 - \lambda \cdot e^{-t(\mu-\lambda)} \cdot \frac{1-z}{\mu-\lambda z}} \right]^i = \\ &= \left[\frac{\mu - \lambda z - \mu e^{-t(\mu-\lambda)}(1-z)}{\mu - \lambda z - \lambda e^{-t(\mu-\lambda)}(1-z)} \right]^i = \\ &= \left[\frac{\mu(1 - e^{-t(\mu-\lambda)}) - (\lambda - \mu e^{-t(\mu-\lambda)})z}{\mu - \lambda e^{-t(\mu-\lambda)} - \lambda(1 - e^{-t(\mu-\lambda)})z} \right]^i = \\ &= \left[\frac{\frac{\mu(1 - e^{-t(\mu-\lambda)})}{\mu - \lambda e^{-t(\mu-\lambda)}} - \frac{\lambda - \mu e^{-t(\mu-\lambda)}}{\mu - \lambda e^{-t(\mu-\lambda)}} z}{1 - \frac{\lambda(1 - e^{-t(\mu-\lambda)})}{\mu - \lambda e^{-t(\mu-\lambda)}} z} \right]^i = \\ &= \left[\frac{r_t - (r_t + q_t - 1)z}{1 - q_t z} \right]^i = \\ &= \left[\frac{r_t - r_t(q_t + p_t)z + p_t z}{1 - q_t z} \right]^i = \left[r_t + (1 - r_t) \frac{p_t z}{1 - q_t z} \right]^i \end{aligned}$$

which is a **composition** of a Binomial distribution with i trials and $1 - r_t$ for the probability of a success, and a Geometric distribution for which the probability of a success is p_t .

Extension

to solve a slightly more complicated

$$\dot{P}(z, t) = a(z) \cdot P'(z, t) + b(z) \cdot P(z, t)$$

One can show that the general solution can be found by solving the 'homogenous' version of this equation first (which we already know how to do - I'll now call it

$G(z, t)$), and then multiplying it by a function of z , say $h(z)$, substituting into the full (non-homogeneous) equation, and solving for $h(z)$. This yields

$$\begin{aligned} \dot{G}(z, t)h(z) = \\ a(z)G'(z, t)h(z) + a(z)G(z, t)h'(z) + b(z)G(z, t), h(z) \end{aligned}$$

Since

$$\dot{G}(z, t) = a(z)G'(z, t)$$

we can cancel the first two terms and write

$$0 = a(z)G(z, t)h'(z) + b(z)G(z, t), h(z)$$

or

$$\frac{h'(z)}{h(z)} = -\frac{b(z)}{a(z)}$$

which can be easily solved:

$$h(z) = \exp\left(-\int \frac{b(z)}{a(z)} dz\right)$$

To solve the initial-value problem, we now have to find $g(\cdot)$ such that

$$g\left[\exp\left(-\int \frac{dz}{a(z)}\right)\right] \cdot h(z) = z^i$$

Examples:

1. Linear Growth with immigration:

$$\begin{aligned} a(z) &= (1-z)(\mu - \lambda z) \\ b(z) &= -a(1-z) \end{aligned}$$

We have already solved the homogeneous version, so let us find

$$\begin{aligned} h(z) &= \exp\left(a \int \frac{1-z}{(1-z)(\mu-\lambda z)} dz\right) = \\ &= \exp\left(-\frac{a}{\lambda} \ln(\mu - \lambda z)\right) = (\mu - \lambda z)^{-a/\lambda} \end{aligned}$$

The general solution is thus

$$g\left(e^{-t(\mu-\lambda)} \cdot \frac{1-z}{\mu-\lambda z}\right) (\mu - \lambda z)^{-a/\lambda}$$

We will first solve it assuming that $i = 0$, which implies

$$\begin{aligned} g\left(\frac{1-z}{\mu-\lambda z}\right) (\mu - \lambda z)^{-a/\lambda} &= 1 \\ g\left(\frac{1-z}{\mu-\lambda z}\right) &= (\mu - \lambda z)^{a/\lambda} \\ g(x) &= (\mu - \lambda \frac{1-\mu x}{1-\lambda x})^{a/\lambda} = \left(\frac{\mu-\lambda}{1-\lambda x}\right)^{a/\lambda} \end{aligned}$$

This yields

$$\begin{aligned}
& \left(\frac{\mu - \lambda}{1 - \lambda e^{-t(\mu - \lambda)} \cdot \frac{1-z}{\mu - \lambda z}} \right)^{a/\lambda} (\mu - \lambda z)^{-a/\lambda} = \\
& \left(\frac{\mu - \lambda}{\mu - \lambda z - \lambda e^{-t(\mu - \lambda)}(1-z)} \right)^{a/\lambda} = \\
& \left(\frac{\mu - \lambda}{\mu - \lambda e^{-t(\mu - \lambda)} - \lambda(1 - e^{-t(\mu - \lambda)})z} \right)^{a/\lambda} = \\
& \left(\frac{\frac{\mu - \lambda}{\mu - \lambda e^{-t(\mu - \lambda)}}}{1 - \frac{\lambda(1 - e^{-t(\mu - \lambda)})}{\mu - \lambda e^{-t(\mu - \lambda)}}z} \right)^{a/\lambda} = \left(\frac{p_t}{1 - q_t z} \right)^{a/\lambda}
\end{aligned}$$

which is the *modified* Negative binomial distribution with parameters p_t and $\frac{a}{\lambda}$.

2. M/M/ ∞ queue:

$$\begin{aligned}
a(z) &= \mu(1 - z) \\
b(z) &= -a(1 - z)
\end{aligned}$$

The general solution to the homogeneous version of the equation is

$$\begin{aligned}
g_o \left[e^{-t} \cdot \exp \left(-\frac{1}{\mu} \int \frac{dz}{1-z} \right) \right] &= g_o \left[e^{-t} \cdot (1-z)^{1/\mu} \right] = \\
g \left[e^{-\mu t} \cdot (1-z) \right]
\end{aligned}$$

Then, we get

$$h(z) = \exp \left(\frac{a}{\mu} \int dz \right) = \exp \left(\frac{a}{\mu} z \right)$$

The general solution is thus

$$g \left[e^{-\mu t} \cdot (1-z) \right] \cdot \exp \left(\frac{a}{\mu} z \right)$$

To meet the initial condition, we need

$$g(1-z) \cdot \exp \left(\frac{a}{\mu} z \right) = z^i$$

or

$$g(1-z) = \exp \left(-\frac{a}{\mu} z \right) \cdot z^i$$

implying

$$g(x) = \exp \left[-\frac{a}{\mu}(1-x) \right] \cdot (1-x)^i$$

Finally, we have to replace x by the original argument of g , and multiply by $h(z)$:

$$\begin{aligned}
& \exp \left\{ -\frac{a}{\mu} [1 - e^{-\mu t} \cdot (1-z)] \right\} \cdot [1 - e^{-\mu t} \cdot (1-z)]^i \cdot \exp \left(\frac{a}{\mu} z \right) = \\
& \exp \left\{ -\frac{a}{\mu} (1 - e^{-\mu t} + e^{-\mu t} z - z) \right\} \cdot (q_t + p_t z)^i =
\end{aligned}$$

$$\exp\left[-\frac{\alpha \cdot q_t}{\mu}(1-z)\right] \cdot (q_t + p_t z)^i$$

where $p_t = e^{-\mu t}$. This is an independent sum (convolution) of a Poisson distribution with $\Lambda = \frac{\alpha \cdot q_t}{\mu}$, and a Binomial distribution with i trials and p_t as the probability of a success.

3. Power-supply process (N welders):

$$\begin{aligned} a(z) &= (1-z)(\mu + \lambda z) \\ b(z) &= -N\lambda(1-z) \end{aligned}$$

Since

$$\int \frac{dz}{(1-z)(\mu + \lambda z)} = \frac{1}{\mu + \lambda} \int \left(\frac{1}{1-z} + \frac{\lambda}{\mu + \lambda z} \right) dz = \frac{1}{\lambda + \mu} [\ln(\mu + \lambda z) - \ln(1-z)]$$

the general solution to the homogeneous version of the equation is

$$g_0 \left[e^{-t} \cdot \left(\frac{1-z}{\mu + \lambda z} \right)^{\frac{1}{\lambda + \mu}} \right] = g \left[e^{-t(\lambda + \mu)} \cdot \frac{1-z}{\mu + \lambda z} \right]$$

Then, we get

$$h(z) = \exp\left(N\lambda \int \frac{dz}{\mu + \lambda z}\right) = (\mu + \lambda z)^N$$

The general solution is thus

$$g \left[e^{-t(\lambda + \mu)} \cdot \frac{1-z}{\mu + \lambda z} \right] \cdot (\mu + \lambda z)^N$$

To meet the initial condition, we need

$$g\left(\frac{1-z}{\mu + \lambda z}\right) \cdot (\mu + \lambda z)^N = z^i$$

or

$$g\left(\frac{1-z}{\mu + \lambda z}\right) = z^i \cdot (\mu + \lambda z)^{-N}$$

implying

$$g(x) = \left(\frac{1-\mu x}{1+\lambda x} \right)^i \cdot \left(\frac{\lambda + \mu}{1+\lambda x} \right)^{-N}$$

Finally, we have to replace x by the original argument of g , and multiply

by $h(z)$:

$$\begin{aligned}
& \left(\frac{1 - \mu e^{-t(\lambda+\mu)} \frac{1-z}{\mu+\lambda z}}{1 + \lambda e^{-t(\lambda+\mu)} \frac{1-z}{\mu+\lambda z}} \right)^i \left(\frac{1 + \lambda e^{-t(\lambda+\mu)} \frac{1-z}{\mu+\lambda z}}{\lambda + \mu} \right)^N (\mu + \lambda z)^N = \\
& \left(\frac{\mu + \lambda z - \mu e^{-t(\lambda+\mu)}(1-z)}{\mu + \lambda z + \lambda e^{-t(\lambda+\mu)}(1-z)} \right)^i \cdot \\
& \left(\frac{\mu + \lambda z + \lambda e^{-t(\lambda+\mu)}(1-z)}{\lambda + \mu} \right)^N = \\
& \left(\frac{\mu(1 - e^{-t(\mu+\lambda)}) + (\lambda + \mu e^{-t(\mu+\lambda)})z}{\lambda + \mu} \right)^i \cdot \\
& \left(\frac{\mu + \lambda e^{-t(\mu+\lambda)} + \lambda(1 - e^{-t(\mu+\lambda)})z}{\lambda + \mu} \right)^{N-i} = \\
& \left(q_t^{(1)} + p_t^{(1)} z \right)^i \cdot \left(q_t^{(2)} + p_t^{(2)} z \right)^{N-i}
\end{aligned}$$

(a convolution of two Binomials), where

$$\begin{aligned}
p_t^{(1)} &\equiv \frac{\lambda + \mu e^{-t(\mu+\lambda)}}{\lambda + \mu} \\
p_t^{(2)} &\equiv \frac{\lambda(1 - e^{-t(\mu+\lambda)})}{\lambda + \mu}
\end{aligned}$$