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## A SIZE STRUCTURED MODEL OF THE POPULATION BASED ON A STOCHASTIC GROWTH EQUATION

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**Abstract**. A stochastic growth equation is given whose solution density satisfies size-structured population growth equation and boundary conditions.

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Let  $N(t) = \int_0^\infty u(x,t)dx$  be the decomposition of population by the size. Then in [1],[2] the optimal harvesting (catching) problem for the system

$$\partial_t u(x,t) + \partial_x (g(x)u(x,t)) + m(x)u(x,t) = -\mu(t)u(x,t), \tag{1}$$

$$u(x,0) = u_0(x), \tag{2}$$

$$g(0)u(0,t) = \int_0^\infty \beta(t,x)u(x,t)dx. \tag{3}$$

under revenue function

$$\int_0^T \int_0^\infty \mu(t) u(x,t) dx dt = \int_0^T \mu(t) N(t) dt \xrightarrow{\mu} \max$$

was studied. Here m(x),  $\beta(t,x)$  denote natural mortality and fecundity of population of size x,  $\mu(t)$  denotes fishing mortality and g(x) is rate of growth of individuals.

Our aim is to show that the solution of (1)-(3) may be represented by the distribution function of a stochastic processes.

Let  $\eta_k, k = 1, 2, ...$  be i.i.d. nonnegative random variables with probability density  $f(x), x \geq 0$ ,  $\sigma_n = \sum_k^n \eta_k$  and let  $\eta_0$  be an independent random variable with probability density  $f_0(x)$ . Denote  $N(t) = \#\{n : \eta_0 + \sigma_n \leq t\}$ . Suppose  $A(t) = t - \sigma_{N_t}$  and  $a(t, x) = P(A(t) \leq x)$ .

A) Let  $u(x) = \sum_{n=1}^{\infty} f^{n*}(x)$  and let  $f_0$  be differentiable and  $\tilde{u} = f_0 + u * f$ , where \* denotes convolution.

**Lemma 1.** Let condition A) be satisfied. Then the equation  $E\Phi(A(t)) = \int_0^\infty \Phi(x)(1 - F(x))\tilde{u}(t-x)dx$  is satisfied for every continuous bounded function  $\Phi$ .

*Proof.* For each bounded, continuous function  $\Phi$  we get

$$E\Phi(A_t) = \sum_{n=0}^{\infty} E\Phi(t - \tau_n) I_{(\tau_n \le t < \tau_{n+1})}$$

$$= \sum_{n=0}^{\infty} E\Phi(t - \tau_n) I_{(t_n \le t < \tau_{n+1})} (1 - F(t - \tau_n))$$

$$= \sum_{n=0}^{\infty} \int_0^t \Phi(t - s) (1 - F(t - s)) dF^{*n}(s).$$

$$= \int_0^t \Phi(t - s) (1 - F(t - s)) dU(s) = \int_0^t \Phi(s) (1 - F(s)) u(t - s) ds.$$

Let x(t) be the solution of the growth equation

$$\dot{x}(t) = g(x(t)), x(0) = 0.$$

Let  $Y(t) = \begin{cases} x(t - \tau_{N_t}), & t < \zeta \\ \partial, & t \ge \zeta \end{cases}$ , where  $\zeta = \inf\{s; \int_0^s \mu(v) dv > \tau\}$  and  $\tau$  independent, exponentially distributed r. v. with parameter 1. Then Y satisfies the equation

$$Y(t) = Y(0) + \int_0^t g(Y(s))ds - \int_0^t Y(s-)dN(s), \ t < \zeta.$$

For each bounded, continuous function  $\varphi$  one obtains

$$\int_0^\infty \varphi(r)\rho(t,r)dr = E\varphi(Y(t)), t < \zeta) = E\varphi(A(t)), t < \zeta$$

$$= E\varphi(x(t-\tau_{N_t}))e^{-\int_0^t \mu(v)dv}.$$

Using lemma 1 for the function  $\Phi(t) = \varphi(x(t))$  we get

$$E\varphi(x(t-\tau_{N_t}))e^{-\int_0^t \mu(v)dv} = e^{-\int_0^t \mu(v)dv} \int_0^t \Phi(s)(1-F(s))\tilde{u}(t-s)ds$$

$$= e^{-\int_0^t \mu(v)dv} \int_0^\infty \varphi(x(s))(1-F(s))\tilde{u}(t-s)/g(x(s))dx(s)$$

$$= e^{-\int_0^t \mu(v)dv} \int_0^\infty \varphi(r)(1-F(x^{-1}(r)))\tilde{u}(t-x^{-1}(r))/g(r)dr$$

and 
$$\rho(t,r)e^{\int_0^t \mu(v)dv} = (1 - F(x^{-1}(r)))\tilde{u}(t - x^{-1}(r))/g(r)$$
. Hence 
$$\partial_t \rho(t,r)e^{\int_0^t \mu(v)dv} + \mu(t)\rho(t,r)e^{\int_0^t \mu(v)dv} + \partial_r(g(r)\rho(t,r))e^{\int_0^t \mu(v)dv} + \partial_r(g(r)\rho(t,r))e^{\int_0^t \mu(v)dv}$$
$$= \partial_t(\rho(t,r)e^{\int_0^t \mu(v)dv}) + \partial_r(g(r)\rho(t,r)e^{\int_0^t \mu(v)dv})$$
$$= (1 - F(x^{-1}(r)))\tilde{u}'(t - x^{-1}(r))/g(r)$$
$$-(1 - F(x^{-1}(r)))\tilde{u}'(t - x^{-1}(r))x^{-1'}(r)$$
$$-f(x^{-1}(r)))\tilde{u}(t - x^{-1}(r))x^{-1'}(r)$$
$$= -f(x^{-1}(r)))\tilde{u}(t - x^{-1}(r))/g(r)$$
$$= -\frac{f(x^{-1}(r))}{1 - F(x^{-1}(r))}e^{\int_0^t \mu(v)dv}\rho(t,r).$$

Integrating the equation  $\partial_t \rho(t,r) + \mu(t)\rho(t,r) + \partial_r(g(r)\rho(t,r)) + \frac{f(x^{-1}(r))}{1-F(x^{-1}(r))}\rho(t,r) = 0$  and using  $\int_0^\infty \rho(t,r)dr = e^{-\int_0^t \mu(v)dv}$  one obtains

$$g(0)\rho(t,0) = \int_0^\infty \frac{f(x^{-1}(r))}{1 - F(x^{-1}(r))} \rho(t,r) dr.$$

Finally we get

**Proposition 1.** Let condition A) be satisfied. Then

$$\partial_t \rho(t,r) + \partial_r (g(r)\rho(t,r)) + \mu(t)\rho(t,r) + \frac{f(x^{-1}(r))}{1 - F(x^{-1}(r))}\rho(t,r) = 0,$$

$$g(0)\rho(t,0) = \int_0^\infty \beta(r)\rho(t,r)dr,$$

where 
$$\beta(r) = \frac{1}{1 - F(x^{-1}(r))} (f(x^{-1}(r)).$$

Corollary. The function  $u(x,t) = N_0 \rho(t,x)$  satisfies equations (3).

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