

A Stochastic Model of a Pioneer/Climax Interaction

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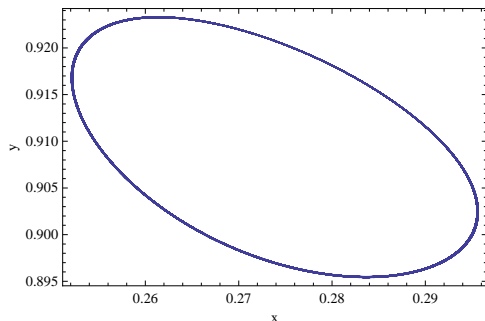
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Deterministic Pioneer/Climax Model

$$\begin{aligned}x'(t) &= f(x(t), y(t)) \\ &= x(t)(1 - c_{11}x(t) - y(t)) \\ y'(t) &= g(x(t), y(t)) \\ &= -y(t)(1 - x(t) - c_{22}y(t)) \left(\frac{9}{8} - x(t) - c_{22}y(t) \right)\end{aligned}$$

- Per capita reproduction rates for pioneer and climax species,
- Hopf bifurcations of equilibria,
- Variables x and y represent population **density** rather than **numbers of individuals**.

Background



- Asymptotic behaviors of pioneer/climax model explored in Buchanan (1999).
- For small populations, the behavior may be described by a stochastic process.
- Approach to the development of a stochastic model is based on the work of Klebaner and Liptser (2001) on the Lotka-Volterra model.

Integral Form of Deterministic Model

$$x(t) - x(0) = \int_0^t f(x(s), y(s)) ds$$
$$y(t) - y(0) = \int_0^t g(x(s), y(s)) ds.$$

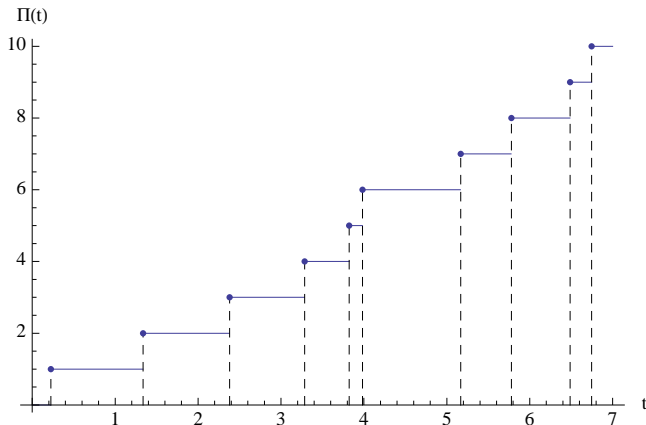
Let K be an integer and define

$$X(0) = Kx(0)$$
$$Y(0) = Ky(0),$$

the initial values of random variables $X(t)$ and $Y(t)$.

Poisson Process

$\Pi'(t)$ is a **Poisson process** with **intensity** (or **rate**) r .



Compensators and Martingales

Remarks:

- Poisson process $\Pi^r(t)$ is monotonic.
- $E[\Pi^r(t)] = rt$.
- The **compensator** of $\Pi^r(t)$ is the deterministic expression rt .
- $\Pi^r(t) - rt$ is a **martingale**, *i.e.* it has no tendency to rise or fall.

$$E[\Pi^r(t) - rt] = 0 \quad \text{for all } t.$$

Stochastic Integral

- $\{\Pi_n^r(t)\}_{n \in \mathbb{N}}$ is a sequence of independent Poisson processes with rate r .
- Let $r [X^i Y^j] (t)$ denote the stochastic integral

$$r [X^i Y^j] (t) = \int_0^t \sum_{n=1}^{\infty} \mathbb{I} \left(X^i(s-) Y^j(s-) \geq n \right) d\Pi_n^r(s). \quad (1)$$

- Define $A_r^{i,j}(t)$ to be

$$\begin{aligned} A_r^{i,j}(t) &= \int_0^t r X^i(s) Y^j(s) ds \\ &= \int_0^t \sum_{n=1}^{\infty} \mathbb{I} \left(X^i(s-) Y^j(s-) \geq n \right) r ds. \end{aligned} \quad (2)$$

State-dependent Intensity of Poisson Processes

$$\begin{aligned} r [X^i Y^j] (t) - A_r^{i,j}(t) \\ = \int_0^t \sum_{n=1}^{\infty} \mathbb{I} \left(X^i(s-) Y^j(s-) \geq n \right) d(\Pi_n^r(s) - rs) \end{aligned}$$

- $r [X^i Y^j] (t) - A_r^{i,j}(t)$ is a local martingale,
- $A_r^{i,j}(t)$ is a compensator for $r [X^i Y^j]_r (t)$,
- the intensity of $r [X^i Y^j] (t)$ is $rX^i(t)Y^j(t)$.

Stochastic Analogue of Deterministic Model

$$\begin{aligned} X(t) = & X(0) + [X Y^0] (t) \\ & - \frac{c_{11}}{K} [X^2 Y^0] (t) - \frac{1}{K} [X Y] (t) \end{aligned} \quad (3)$$

$$\begin{aligned} Y(t) = & Y(0) + \frac{17}{8K} [X Y] (t) + \frac{17c_{22}}{8K} [X^0 Y^2] (t) \\ & - \frac{9}{8} [X^0 Y] (t) - \frac{1}{K^2} [X^2 Y] (t) - \frac{2c_{22}}{K^2} [X Y^2] (t) \\ & - \frac{c_{22}^2}{K^2} [X^0 Y^3] (t). \end{aligned} \quad (4)$$

Blow-up of Solutions

- The stochastic integrals defined in Eq. (1) exist so long as $X(t)$ and $Y(t)$ remain finite.
- The stochastic process $X(t)$ is said to be **honest** if $P(X(t) < \infty) = 1$ for all t , similarly for $Y(t)$.
- Notation:
 - ▶ $a \wedge b = \min\{a, b\}$
 - ▶ $a \vee b = \max\{a, b\}$

Blow-up Time

Define

- $T_\infty = \inf\{t > 0 : X(t) \vee Y(t) = \infty\}$,
- for each $n \in \mathbb{N}$, $T_n^X = \inf\{t > 0 : X(t) \geq n\}$,
- $T_\infty^X = \lim_{n \rightarrow \infty} T_n^X$,
- for each $m \in \mathbb{N}$ define $T_m^Y = \inf\{t > 0 : Y(t) \geq m\}$ and,
- $T_\infty^Y = \lim_{m \rightarrow \infty} T_m^Y$.

Boundedness of $X(t)$ (1 of 3)

$$\begin{aligned} X(t \wedge T_n^X) &= X(0) + [X Y^0] (t \wedge T_n^X) \\ &\quad - \frac{c_{11}}{K} [X^2 Y^0] (t \wedge T_n^X) - \frac{1}{K} [X Y] (t \wedge T_n^X) \\ &\leq X(0) + [X Y^0] (t \wedge T_n^X) \\ &= X(0) + \int_0^{t \wedge T_n^X} \sum_{m=1}^{\infty} \mathbb{I}(X(s-) \geq m) d\Pi_m^1(s). \end{aligned}$$

Boundedness of $X(t)$ (2 of 3)

Taking the expectation of both sides of the inequality yields

$$\begin{aligned} \mathbb{E} \left[X(t \wedge T_n^X) \right] &\leq X(0) + \mathbb{E} \left[\int_0^{t \wedge T_n^X} \sum_{m=1}^{\infty} \mathbb{I}(X(s-) \geq m) d\Pi_m^1(s) \right] \\ &= X(0) + \mathbb{E} \left[\int_0^{t \wedge T_n^X} \sum_{m=1}^{\infty} \mathbb{I}(X(s-) \geq m) ds \right] \\ &= X(0) + \mathbb{E} \left[\int_0^{t \wedge T_n^X} X(s) ds \right] \\ &= X(0) + \mathbb{E} \left[\int_0^{t \wedge T_n^X} X(s \wedge T_n^X) ds \right] \\ &\leq X(0) + \mathbb{E} \left[\int_0^t X(s \wedge T_n^X) ds \right] \\ &= X(0) + \int_0^t \mathbb{E} \left[X(s \wedge T_n^X) \right] ds. \end{aligned}$$

Boundedness of $X(t)$ (3 of 3)

According to Gronwall's inequality

$$\mathbb{E} \left[X(t \wedge T_n^X) \right] \leq X(0)e^t \quad \text{for all } n \in \mathbb{N}, t > 0.$$

Fatou's Lemma implies

$$\begin{aligned} \mathbb{E} \left[X(t \wedge T_\infty^X) \right] &= \mathbb{E} \left[\lim_{n \rightarrow \infty} X(t \wedge T_n^X) \right] \\ &\leq \lim_{n \rightarrow \infty} \mathbb{E} \left[X(t \wedge T_n^X) \right] \\ &\leq X(0)e^t. \end{aligned}$$

It follows that $\mathbb{P}(T_\infty^X \leq t) = 0$ for all $t > 0$, so $X(t)$ is honest.

Boundedness of $Y(t)$ (1 of 4)

$$\begin{aligned} & Y(t \wedge T_n^X \wedge T_m^Y) \\ & \leq Y(0) + \frac{17}{8K} [X Y](t \wedge T_n^X \wedge T_m^Y) + \frac{17c_{22}}{8K} [X^0 Y^2](t \wedge T_n^X \wedge T_m^Y) \\ & \quad - \frac{2c_{22}}{K^2} [X Y^2](t \wedge T_n^X \wedge T_m^Y) \\ & = Y(0) + \int_0^{t \wedge T_n^X \wedge T_m^Y} \sum_{j=1}^{\infty} \mathbb{I}(X(s-)Y(s-) \geq j) d\pi_j^{17/8K}(s) \\ & \quad + \int_0^{t \wedge T_n^X \wedge T_m^Y} \sum_{j=1}^{\infty} \mathbb{I}(Y^2(s-) \geq j) d\pi_j^{17c_{22}/8K}(s) \\ & \quad - \int_0^{t \wedge T_n^X \wedge T_m^Y} \sum_{j=1}^{\infty} \mathbb{I}(X(s-)Y^2(s-) \geq j) d\pi_j^{2c_{22}/K^2}(s). \end{aligned}$$

Boundedness of $Y(t)$ (2 of 4)

Taking expectations of both sides of the inequality above produces

$$\begin{aligned} & \mathbb{E} \left[Y(t \wedge T_n^X \wedge T_m^Y) \right] \\ & \leq Y(0) + \int_0^t \frac{17}{8K} \mathbb{E} \left[X(s \wedge T_n^X \wedge T_m^Y) Y(s \wedge T_n^X \wedge T_m^Y) \right] ds \\ & \quad + \int_0^t \frac{17c_{22}}{8K} \mathbb{E} \left[Y^2(s \wedge T_n^X \wedge T_m^Y) \right] ds \\ & \quad - \int_0^t \frac{2c_{22}}{K^2} \mathbb{E} \left[X(s \wedge T_n^X \wedge T_m^Y) Y^2(s \wedge T_n^X \wedge T_m^Y) \right] ds \\ & \leq Y(0) + \int_0^t \frac{17n}{8K} \mathbb{E} \left[Y(s \wedge T_n^X \wedge T_m^Y) \right] ds \\ & \quad + \int_0^t \frac{17c_{22}}{8K} \mathbb{E} \left[Y^2(s \wedge T_n^X \wedge T_m^Y) \right] ds \\ & \quad - \int_0^t \frac{2nc_{22}}{K^2} \mathbb{E} \left[Y^2(s \wedge T_n^X \wedge T_m^Y) \right] ds \end{aligned}$$

Boundedness of $Y(t)$ (3 of 4)

$$\begin{aligned} & \mathbb{E} \left[Y(t \wedge T_n^X \wedge T_m^Y) \right] \\ & \leq Y(0) + \int_0^t \frac{17n}{8K} \mathbb{E} \left[Y(s \wedge T_n^X \wedge T_m^Y) \right] ds \\ & \quad + \frac{c_{22}}{K} \left(\frac{17}{8} - \frac{2n}{K} \right) \int_0^t \mathbb{E} \left[Y^2(s \wedge T_n^X \wedge T_m^Y) \right] ds \\ & \leq Y(0) + \frac{17n}{8K} \int_0^t \mathbb{E} \left[Y(s \wedge T_n^X \wedge T_m^Y) \right] ds \end{aligned}$$

for $n \geq 17K/16$.

Boundedness of $Y(t)$ (4 of 4)

Gronwall's inequality implies

$$\mathbb{E} \left[Y(t \wedge T_n^X \wedge T_m^Y) \right] \leq Y(0) e^{17nt/(8K)}$$

for all $t > 0$ when $n \geq 17K/16$.

Fatou's Lemma implies

$$\mathbb{E} \left[Y(t \wedge T_n^X \wedge T_\infty^Y) \right] \leq Y(0) e^{17nt/(8K)}.$$

Therefore $\mathbb{P}(T_\infty^Y \leq T_n^X \wedge t) = 0$ for all $t > 0$ when $n \geq 17K/16$.
Since $T_n^X \rightarrow \infty$ as $n \rightarrow \infty$ then

$$\lim_{n \rightarrow \infty} (T_n^X \wedge t) = t$$

for all $t > 0$ and thus $\mathbb{P}(T_m^Y \leq t) = 0$.

Boundedness of $(X(t), Y(t))$

Since $T_\infty = T_\infty^X \wedge T_\infty^Y$ then

$$\begin{aligned} \mathbf{P}(T_\infty \leq t) &= \mathbf{P}(T_\infty^X \wedge T_\infty^Y \leq t) \\ &\leq \mathbf{P}(T_\infty^X \leq t) + \mathbf{P}(T_\infty^Y \leq t) \\ &= 0. \end{aligned}$$

Quadratic Variation and Martingales

Define

$$M_r^{i,j}(t) = r \left[X^i Y^j \right] (t) - A_r^{i,j}(t)$$

which is a **local martingale**.

The **quadratic variation process** of $N(t)$ is defined as

$$[N, N](t) = \lim_{\|P\| \rightarrow 0} \sum_{i=0}^{n-1} (N(t_{i+1}^n) - N(t_i^n))^2$$

where $\{t_i^n\}_{i=0}^n$ is any partition of $[0, t]$.

The **sharp bracket process** is the compensator of $[N, N](t)$.

$$\left\langle M_r^{i,j}(t), M_r^{i,j}(t) \right\rangle = A_r^{i,j}(t)$$

Normalized Stochastic Random Variables

Define:

$$x^K(t) = \frac{X(t)}{K}$$

$$y^K(t) = \frac{Y(t)}{K}$$

$$m(t) = \frac{1}{K} \left(M_1^{1,0}(t) - M_{\frac{c_{11}}{K}}^{2,0}(t) - M_{\frac{1}{K}}^{1,1}(t) \right)$$

$$\hat{m}(t) = \frac{1}{K} \left(M_{\frac{17}{8K}}^{1,1}(t) - M_{\frac{17c_{22}}{8K}}^{0,2}(t) - M_{\frac{9}{8}}^{0,1}(t) - M_{\frac{1}{K^2}}^{2,1}(t) \right. \\ \left. - M_{\frac{2c_{22}}{K^2}}^{1,2}(t) - M_{\frac{c_{22}^2}{K^2}}^{0,3}(t) \right)$$

Stochastic Integral Equations

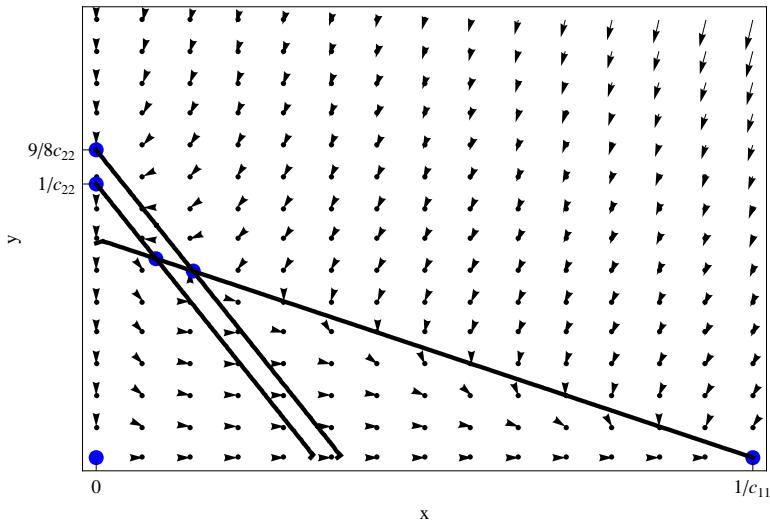
Dividing Eqs. (3) and (4) by K produces a pair of integral equations for the stochastic process $(x^K(t), y^K(t))$.

$$x^K(t) - x(0) = \int_0^t f(x^K(s), y^K(s)) ds + m(t)$$

$$y^K(t) - y(0) = \int_0^t g(x^K(s), y^K(s)) ds + \hat{m}(t)$$

Limiting Case as $K \rightarrow \infty$

Question: does the behavior of $(x^K(t), y^K(t))$ resemble the behavior of $(x(t), y(t))$ as $K \rightarrow \infty$?



Difference of Solutions

Define:

$$\begin{aligned}T_n &= \inf\{t > 0 : X(t) \vee Y(t) \geq n\} \\T_n^K &= \inf\{t > 0 : x^K(t) \vee y^K(t) \geq n\}\end{aligned}$$

Note: $T_n^K = T_{nK}$

Fix $T > 0$ and $n \in \mathbb{N}$ and consider the differences

$$f(x^K(t), y^K(t)) - f(x(t), y(t)) \quad \text{and} \quad g(x^K(t), y^K(t)) - g(x(t), y(t)),$$

for $0 \leq t \leq T_n^K \wedge T$.

Deviation of Solutions

There exists constant $L_n > 0$ depending on n and T such that

$$\begin{aligned} & \left| f(x^K(t), y^K(t)) - f(x(t), y(t)) \right| \\ & \leq L_n \left(\left| x^K(t) - x(t) \right| + \left| y^K(t) - y(t) \right| \right) \\ & \left| g(x^K(t), y^K(t)) - g(x(t), y(t)) \right| \\ & \leq L_n \left(\left| x^K(t) - x(t) \right| + \left| y^K(t) - y(t) \right| \right), \end{aligned}$$

for $0 \leq t \leq T_n^K \wedge T$, which implies

$$\begin{aligned} & \left| x^K(t) - x(t) \right| + \left| y^K(t) - y(t) \right| \\ & \leq 2L_n \int_0^t \left(\left| x^K(s) - x(s) \right| + \left| y^K(s) - y^K(s) \right| \right) ds \\ & \quad + |m(t)| + |\hat{m}(t)|. \end{aligned}$$

Supremum of the Deviation

$$\begin{aligned} & \sup_{t \leq T \wedge T_n^K} \left(\left| x^K(t) - x(t) \right| + \left| y^K(t) - y(t) \right| \right) \\ & \leq 2L_n \int_0^t \sup_{s \leq T \wedge T_n^K} \left(\left| x^K(s) - x(s) \right| + \left| y^K(s) - y^K(s) \right| \right) ds \\ & \quad + \sup_{t \leq T \wedge T_n^K} |m(t)| + \sup_{t \leq T \wedge T_n^K} |\hat{m}(t)| \end{aligned}$$

Gronwall's inequality implies

$$\begin{aligned} & \sup_{t \leq T \wedge T_n^K} \left(\left| x^K(t) - x(t) \right| + \left| y^K(t) - y(t) \right| \right) \\ & \leq \left(\sup_{t \leq T \wedge T_n^K} |m(t)| + \sup_{t \leq T \wedge T_n^K} |\hat{m}(t)| \right) e^{2L_n T}. \end{aligned} \quad (5)$$

Bounding the Normalized Martingales (1 of 4)

Compute the sharp brackets:

$$\langle m, m \rangle (t) = \left\langle \frac{1}{K} \left(M_1^{1,0} - M_{c_{11}/K}^{2,0} - M_{1/K}^{1,1} \right), \frac{1}{K} \left(M_1^{1,0} - M_{c_{11}/K}^{2,0} - M_{1/K}^{1,1} \right) \right\rangle (t)$$

$$= \frac{1}{K} \int_0^t x^K(s) \left(1 + c_{11} x^K(s) + y^K(s) \right) ds$$

$$\langle \hat{m}, \hat{m} \rangle (t)$$

$$= \frac{1}{K} \int_0^t y^K(s) \left(1 + x^K(s) + c_{22} y^K(s) \right) \left(\frac{9}{8} + x^K(s) + c_{22} y^K(s) \right) ds$$

The bilinearity of the sharp bracket and the independence of the jumps in the state-dependent Poisson random variables has been used.

Bounding the Normalized Martingales (2 of 4)

By definition of T_n^K ,

$$\sup_{t \leq T \wedge T_n^K} (x^K(t) \vee y^K(t)) \leq n + 1$$

which implies

$$\begin{aligned} & \mathbb{E} \left[\langle m, m \rangle (T \wedge T_n^K) \right] \\ &= \frac{1}{K} \mathbb{E} \left[\int_0^{T \wedge T_n^K} x^K(s) \left(1 + c_{11} x^K(s) + y^K(s) \right) ds \right] \\ &\leq \frac{T}{K} (n + 1) (1 + (c_{11} + 1)(n + 1)) \\ & \mathbb{E} \left[\langle \hat{m}, \hat{m} \rangle (T \wedge T_n^K) \right] \\ &\leq \frac{T}{K} (n + 1) (1 + (c_{22} + 1)(n + 1)) \left(\frac{9}{8} + (c_{22} + 1)(n + 1) \right). \end{aligned}$$

Bounding the Normalized Martingales (3 of 4)

Applying Doob's martingale inequality twice yields

$$\begin{aligned} \mathbb{E} \left[\left(\sup_{t \leq T \wedge T_n^K} m(t) \right)^2 \right] &\leq 4\mathbb{E} \left[\langle m, m \rangle (T \wedge T_n^K) \right] \quad \text{and} \\ \mathbb{E} \left[\left(\sup_{t \leq T \wedge T_n^K} \hat{m}(t) \right)^2 \right] &\leq 4\mathbb{E} \left[\langle \hat{m}, \hat{m} \rangle (T \wedge T_n^K) \right]. \end{aligned}$$

This implies

$$\begin{aligned} \lim_{K \rightarrow \infty} \mathbb{E} \left[\left(\sup_{t \leq T \wedge T_n^K} m(t) \right)^2 \right] &= \lim_{K \rightarrow \infty} \frac{4T}{K} (n+1) (1 + (c_{11} + 1)(n+1)) \\ &= 0 \quad \text{and likewise} \\ \lim_{K \rightarrow \infty} \mathbb{E} \left[\left(\sup_{t \leq T \wedge T_n^K} \hat{m}(t) \right)^2 \right] &= 0 \end{aligned}$$

Bounding the Normalized Martingales (4 of 4)

The limits on the previous slide confirm

$$\lim_{K \rightarrow \infty} \left(\sup_{t \leq T \wedge T_n^K} m(t) + \sup_{t \leq T \wedge T_n^K} \hat{m}(t) \right) = 0. \quad (6)$$

Fluid Approximation

If $\epsilon > 0$ then by Eq. (6) and inequality (5) it follows that

$$\lim_{K \rightarrow \infty} \mathbb{P} \left(\sup_{t \leq T \wedge T_n^K} \left(|x^K(t) - x(t)| + |y^K(t) - y(t)| \right) > \epsilon \right) = 0.$$

If $T > 0$,

$$\lim_{n \rightarrow \infty} \mathbb{P}(T_n \leq T) = 0 \implies \lim_{n \rightarrow \infty} \limsup_{K \rightarrow \infty} \mathbb{P}(T_n^K \leq T) = 0.$$

Hence with almost certainty, $T_n^K \wedge T = T$ and thus for all $\epsilon > 0$

$$\lim_{K \rightarrow \infty} \mathbb{P} \left(\sup_{t \leq T} \left(|x^K(t) - x(t)| + |y^K(t) - y(t)| \right) > \epsilon \right) = 0.$$

Summary

- The deterministic model is the limiting result of the stochastic model for large populations.
- Future work: explore the possibility of extinction of one or both of the species for small to moderately sized populations.