

THE APPLICATION OF STOCHASTIC DIFFERENTIAL EQUATIONS TO POPULATION GENETIC MODEL

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ABSTRACT. In multi-allelic model $X = (x_1, x_2, \dots, x_d)$,

$$M_f(t) = f(p(t)) - \int_0^t Lf(p(t))ds$$

is a P -martingale for diffusion operator L under the certain conditions. In this note, we examine the stochastic differential equation for model X and find the properties using stochastic differential equation.

1. Introduction

Consider a partition to be a sequence

$$X = (x_1, x_2, \dots, x_d) \in R^d.$$

If the partition X has α_i parts equal to i , then we write

$$X = [1^{\alpha_1}, 2^{\alpha_2}, \dots, n^{\alpha_n}].$$

We consider n locus model, so we find n genes on a chromosome. A partition X describes a state of a chromosome and X means that there exist d kinds of alleles which occupy x_1 loci, x_2 loci, \dots , x_d loci. In other words, X describes that there exists α_i kinds of alleles occurring i loci for each i . Let q_{ij} denote “mutation rate” or “gene conversion rate” from a partition X_i to another partition X_j per generation measured on the t time scale and p_i denotes the frequency of type X_i .

In population genetics theory we often encounter diffusion process on the compact domain

$$K = \{(p_1, p_2, \dots, p_d); p_1 \geq 0, \dots, p_d \geq 0, p_1 + p_2 + \dots + p_d = 1\}$$

Received March 10, 2003.

2000 Mathematics Subject Classification: 92D10, 60H30, 60G44.

Key words and phrases: allelic model, martingale problem, stochastic differential equation, random genetic drift, mutation rate, natural selection.

We suppose that the vector $p(t) = (p_1, p_2, \dots, p_d)$ of gene frequencies varies with time t .

Let L be a second order differential operator on K

$$(1.1) \quad L = \sum_{i,j=1}^d a_{ij}(p) \frac{\partial^2}{\partial p_i \partial p_j} + \sum_{i=1}^d b_i(p) \frac{\partial}{\partial p_i}$$

with domain $C^2(K)$, where $\{a_{ij}\}$ is a real symmetric and non-negative definite matrix defined on K and $\{b_i\}$ is an R^d -valued measurable function defined on K . The coefficient $\{a_{ij}\}$ comes from chance replacement of individuals by new ones after random mating and $\{b_i\}$ is represented by the addition of "mutation or gene conversion rate" and the effect of natural selection. The operator L has the same form as the generator of the diffusion describing a $p(t)$ -allele model incorporating mutation and random drift with single locus, but we could give a remark that the matrix q_{ij} depends on the combinatorial structure of the partitions.

We assume that $\{a_{ij}\}$ and $\{b_i\}$ are continuous on K . Let $\Omega = C([0, \infty) : K)$ be the space of all K -valued continuous function defined on $[0, \infty)$. A probability P on (Ω, \mathcal{F}) is called a solution of the (K, L, p) -martingale problem if it satisfies the following conditions,

- (1) $P(p(0) = p) = 1$.
- (2) denoting $M_f(t) = f(p(t)) - \int_0^t Lf(p(s))ds$, $(M_f(t), \mathcal{F}_t)$ is a P -martingale for each $f \in C^2(K)$.

The diffusion operator L was first introduced by Gillespie ([4]) in case that the partition consists of two points. In this case, L is a one-dimensional diffusion operator. However, the uniqueness of solutions of the (K, L, p) -martingale problem has not been generally established. For this problem, Either ([2]) proved that if $\{a_{ij}(p)\} = \{p_i(\delta_{ij} - p_j)\}$ for Kronecker symbol δ_{ij} and $\{b_i(p)\}$ are C^4 -functions satisfying a certain condition, then the uniqueness of the (K, L, p) -martingale problem holds. Also, Okada ([5]) showed that the uniqueness holds for a rather general class in two dimension. In case that L reduces to an infinite allelic diffusion model of the Wright-Fisher type, Either ([3]) gave a partial result.

In this note, we try to apply the stochastic differential equation for multi-allelic model. A key point is that the (K, L, p) -martingale problem in population genetics model is related to simpler stochastic differential equation, so we can find various properties for multi-allelic model. Also, we find a solution of stochastic differential equation for simple-allelic model.

2. Main results

A finite sequence $\{X_1, X_2, \dots, X_K\}$ of partitions is called (X_1, X_K) -chain if X_{i+1} is a consequent of X_i by mutation or gene conversion for each $i = 1, 2, \dots, K - 1$.

We begin with the following Theorem.

THEOREM 1.

$$\rho = \left(\frac{q_{12}}{q_{21}}\right) \left(\frac{q_{23}}{q_{32}}\right) \dots \left(\frac{q_{K-1\ K}}{q_{K\ K-1}}\right) = 1.$$

In other word, the value ρ does not depend on the choice of (X_1, X_K) -chain.

Proof. For simplicity, we consider only a case $K = 3$ and assume X_1 can be changed to X_2 and X_3 by mutation (i.e. X_1 is a consequent of X_2 and X_3 by gene conversion), besides X_2 and X_3 can be changed to each other by gene conversion. Putting

$$\begin{aligned} X_1 &= [1^{\alpha_1}, \dots, i^{\alpha_i}, \dots, j^{\alpha_j}, \dots, n^{\alpha_n}], \\ X_2 &= [1^{\alpha_1+1}, \dots, (i-1)^{\alpha_{i-1}+1}, i^{\alpha_i-1}, \dots, j^{\alpha_j}, \dots, n^{\alpha_n}], \\ X_3 &= [1^{\alpha_1+1}, \dots, i^{\alpha_i}, \dots, (j-1)^{\alpha_{j-1}+1}, j^{\alpha_j-1}, \dots, n^{\alpha_n}], \end{aligned}$$

we have

$$\begin{aligned} q_{12} &= i\alpha_i, \\ q_{21} &= (\alpha_1 + 1)(i - 1)(\alpha_{i-1} + 1), \\ q_{23} &= (i - 1)(\alpha_{i-1} + 1)j\alpha_j, \\ q_{32} &= i\alpha_i(j - 1)(\alpha_{j-1} + 1), \\ q_{31} &= (\alpha_1 + 1)(j - 1)(\alpha_{j-1} + 1), \\ q_{13} &= j\alpha_j. \end{aligned}$$

Obviously, $\rho = 1$. □

Now, we are concerned with diffusion processes associated with second order differential operator L with random genetic drift

$$a_{ij} = p_i\beta_i\delta_{ij} + p_i p_j \left(\sum_{k=1}^d p_k\beta_k - \beta_i - \beta_j\right).$$

Here $\{\beta_i\}$ is non-negative constant satisfying that $\sup_i \beta_i < +\infty$, and δ_{ij} stands for the Kronecker symbol.

We used instead of K the simplex

$$K_1 = \{(p_1, p_2, \dots, p_{d-1}); p_1 \geq 0, \dots, p_{d-1} \geq 0, p_1 + p_2 + \dots + p_{d-1} \leq 1\}.$$

We will restate in the stochastic differential equation, by projecting K onto K_1 .

THEOREM 2. *Suppose that $b_i(p) \geq 0$ whenever $p_i = 0$, and $\sum_{i=1}^d b_i(p) = 0$ for all $p \in K$. Then*

$$\begin{aligned} \langle M_f(t) \rangle &= \sum_{i,j=1}^d \int_0^t (p_i(v)\beta_i(v)\delta_{ij} \\ &\quad + p_i(v)p_j(v)(\sum_{k=1}^d p_i(v)\beta_k(v) - \beta_i(v) - \beta_j(v))) \\ &\quad \times \frac{\partial p(v)}{\partial p_i} \frac{\partial p(v)}{\partial p_j} dv \end{aligned}$$

where $\langle \cdot \rangle$ denotes a quadratic variation process.

Proof. We first choose $\{\alpha_{ij}(p)\}$ as follows :

$$\alpha_{ij}(p) = (\delta_{ij} - p_i)\sqrt{\beta_j p_j}.$$

Then it satisfies

$$(2.1) \quad \sum_{k=1}^{d-1} \alpha_{ik}(p)\alpha_{jk}(p) = p_i\beta_i\delta_{ij} + p_i p_j (\sum_{k=1}^{d-1} p_i\beta_k - \beta_i - \beta_j).$$

Let $p(t)$ be the solution to stochastic differential equation

$$(2.2) \quad dp_i(t) = \sum_{k=1}^{d-1} \alpha_{ik}(p(t))dB_k(t) + b_i(p(t))dt, \quad i = 1, 2, \dots, d-1$$

where $p_d = 1 - (p_1 + p_2 + \dots + p_{d-1})$ and B_1, B_2, \dots, B_{d-1} are independent Brownian motions.

It is well-known that to show the existence and uniqueness of solutions of the (K_1, L, p) -martingale problem is equivalent to show that the stochastic differential equation has a unique solution ([1], [6]).

Let

$$\alpha_{dk} = - \sum_{i=1}^{d-1} \alpha_{ik}.$$

Since $b_d = -\sum_{i=1}^{d-1} b_i$, stochastic differential equation (2.2) holds for $i = d$. Moreover (2.1) holds when $i = d$ or $j = d$. From the Ito stochastic differential rule ([1], [6]),

$$M_f(t) = \sum_{k=1}^{d-1} \int_0^t \theta_k dB_k, \quad \theta_k = \sum_{j=1}^d \frac{\partial \alpha_{jk}}{\partial p_j}.$$

Let $\langle M_f(t) \rangle$ denote the increasing process associated with the martingale $M_f(t)$. Since it has the property that $\langle M_f(0) \rangle = 0$ and $M_f^2(t) - \langle M_f(t) \rangle$ is a martingale [1], [6], we have

$$\begin{aligned} \langle M_f(t) \rangle &= \sum_{k=1}^{d-1} \int_0^t \theta_k^2(v) dv, \\ \langle M_f(t) \rangle &= \sum_{i,j=1}^d \int_0^t \left(p_i(v) \beta_i(v) \delta_{ij} \right. \\ &\quad \left. + p_i(v) p_j(v) \left(\sum_{k=1}^d p_i(v) \beta_k(v) - \beta_i(v) - \beta_j(v) \right) \right) \\ &\quad \times \frac{\partial p(v)}{\partial p_i} \frac{\partial p(v)}{\partial p_j} dv. \end{aligned}$$

□

COROLLARY 3. *Let $a_{ij}(p)$ be any polynomial satisfying (1.1). Suppose that $a_{ii}(p) = 0$ if $p_i = 0$, and $\sum_{i,j=1}^d a_{ij}(p) = 0$ if $\sum_{i=1}^d p_i = 1$. Then, there exists a positive constant ε such that $\langle M_f^\varepsilon(0) \rangle = 0$ and $\{M_f^\varepsilon(t)\}^2 - \langle M_f^\varepsilon(t) \rangle$ is a martingale. Here*

$$\begin{aligned} \langle M_f^\varepsilon(t) \rangle &= \sum_{i,j=1}^d \int_0^t \left(p_i(v) \beta_i(v) \delta_{ij} \right. \\ &\quad \left. + p_i(v) p_j(v) \left(\sum_{k=1}^d p_i(v) \beta_k(v) - \beta_i(v) - \beta_j(v) \right) \right. \\ &\quad \left. + \varepsilon a_{ij}(p) \right) \frac{\partial p(v)}{\partial p_i} \frac{\partial p(v)}{\partial p_j} dv. \end{aligned}$$

Proof. This result follows directly from Theorem 2 and T. Shiga ([7]). □

EXAMPLE. (A model with natural selection) Let the coefficients $\{b_i(p)\}$ have the form

$$b_i(p) = g_i(p) + h_i(p)$$

where

$$g_i(p) = \sum_{k=1}^d p_k q_{ki}$$

$$q_{ij} \geq 0 \text{ for } i \neq j, \quad q_{ii} = - \sum_{j \neq i} q_{ij}.$$

For $i \neq j$, q_{ij} represents a rate of change of type x_i to type x_j .

For a finite number of types x_1, x_2, \dots, x_d , a standard model for incorporating natural selection is to take

$$h_i(p) = p_i \left(\sum_{k=1}^d p_k m_{ki} - \sum_{j,l=1}^d p_j p_l m_{jl} \right).$$

The interpretation is that x_1, x_2, \dots, x_d are possible alleles carried by a gamete at some gene locus, and m_{ij} is a fitness coefficient of the genotype (x_i, x_j) such that $m_{ji} = m_{ij}$. If p_i is the frequency of type i , then $p_i p_j$ represents the frequency of (x_i, x_j) .

Consider a model with natural selection and let $d = 1$. Suppose that $\alpha' = 0$. Then the existence and uniqueness of solutions of the (K, L, p) -martingale problem is equivalent to show that the stochastic differential equation

$$dp_i(t) = \alpha(p_i(t)) \circ dB(t) + [g(p_i(t)) + h(p_i(t))] dt, \quad p_i(0) = p_0$$

has a unique solution. Here operation \circ denotes the symmetric \mathcal{Q} -multiplicaton.

Let $u(z_1, z_2)$ be the solution of

$$\frac{\partial u}{\partial z_2} = a(u), \quad u(z_1, 0) = z_1,$$

and A_t be the solution of

$$\frac{dA_t}{dt} = g(u(A_t, B_t)) + h(u(A_t, B_t)), \quad A_0 = p_0.$$

Then the solution $p_i(t)$ is given by

$$p_i(t) = u(A_t, B_t).$$

Indeed, by the chain rule

$$dp_i(t) = a(u(A_t, B_t)) \circ dB_t + \left(\frac{\partial u}{\partial z_1} \right) (A_t, B_t) (g(u(A_t, B_t)) + h(u(A_t, B_t))) dt,$$

and hence

$$dp_i(t) = \alpha(p_i(t)) \circ dB(t) + [g(p_i(t)) + h(p_i(t))] dt.$$

ACKNOWLEDGEMENT. The first author is grateful to Professor Jian Huang for several helpful suggestions concerning biological ideas during my staying at University of Iowa.

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