

diffusion, a prerequisite introduction to the Langevin diffusion and its associated Poisson's equation is given. The intention of Langevin, that in the special case of Langevin diffusion, the –L algorithm provides a simple and elegant solution. This is a joint property of its differential generator. This is the major contribution in this chapter. A more general version of the algorithm for diffusion processes is also derived along the lines of the standard LSTD algorithm. This algorithm with its application to time analog of the algorithm has been successfully applied to problems in optimal control.

Langevin in the context of MCMC algorithms.

$$d_t = \underbrace{-\nabla U(t)}_{drift} dt + \underbrace{\sqrt{2} W_t}_{diffusion},$$

(1)

$$\{W_t : t \geq 0\}$$

$$\mathbb{R}^d \rightarrow \mathbb{R}$$

$$e^{-\Lambda}$$

$$\Delta$$

Langevin can be thought of as composed of a deterministic drift term and a stochastic diffusion term. The intuition is that the Mauryama scheme are used: $diff_t d_{Langevin} discrete :_n =_{n-1}$

$$-\nabla U(n-1)_n + \sqrt{2} W_{n-1}, (2)$$

$$\{n\}_{n \geq 1}$$

$$\{W_n\}_{n \geq 1}$$

$$n \equiv$$

$$f := \lim_{t \rightarrow 0} \frac{[f(t) - f(x)|_0 = x]}{t} = -\nabla \cdot \nabla f + \Delta f, f \in C^2,$$

(3)

$$\sum_{t \geq 0} P_t - I.$$

$$\mathbb{R} \rightarrow$$

$$\eta := \sim [c()] = \int_c (x)(x)x = \langle c, 1 \rangle_{L^2}.$$

(5)

$$h \in C^2$$

$$h := -, = c - \eta.$$

(6)

$$h(x) = \int_0^\infty x[(t)]t,$$

(7)

$$h(x)$$

$$x[(t)]$$

$$[(t)|_0 =$$

$$x]$$

$$h$$

$$h$$

$$h$$

$$h$$

$$h \in C^2$$

$$h$$

$$h$$

$$h$$

$$h$$

poisson is crucial in each of the applications we consider in this dissertation. In the feedback particle filter (FPF), the innovation

$$\nabla h(x), x \in$$

(9)