

## Problem Set 5 Math 271C, Spring 2020.

1. a) Let  $l_T^x$  be local time (at origin) at time  $T$  for  $Y_t = \sigma\beta_t + x$  with  $\beta$  standard Brownian motion. Set up problems for determining  $\mathbb{E}[l_T^x]$  and  $\mathbb{E}[(l_T^x)^2]$  (the pdes).  
 b) Solve the problem for  $\mathbb{E}[l_T^x]$ .
2. a) Show that the diffusion matrix  $a$  with

$$a_{i,j}(\mathbf{x}) = \int_{-\infty}^{\infty} \mathbb{E}[F_i(\mathbf{x}, Y(0))F_j(\mathbf{x}, Y(t))]dt$$

is non-negative definite (for  $Y$  and ergodic Markov process with notation as in (last) class and notes on Canvas).

- b) Consider the reduced SIR model:

$$\frac{d}{dt}I_t = \kappa I_t(1 - I_t) - \lambda I_t,$$

with the model for the contact rate and the death/recovery rate:

$$\begin{aligned}\kappa &= \kappa_t = \bar{\kappa} + F(Y_{t/\varepsilon}), \\ \lambda &= \bar{\lambda} + \varepsilon\tilde{\lambda},\end{aligned}$$

with  $\varepsilon \ll 1$  and with  $|F| < \bar{\kappa}$  and  $F$  centered w.r.t. the invariant distribution for the ergodic diffusion  $Y$  and where the contact rate fluctuations are relatively rapid (happens on a fast, say daily, scale relative to the recovery/death time scale  $1/\bar{\lambda}$ ). Introduce the time and amplitude rescaling  $t' = \varepsilon^a t$ ,  $I_0 = \varepsilon^b \bar{I}_0$  and identify values for  $a, b, \bar{\lambda}$  so that the problem is in diffusion limit form.

- c) Identify the (weak) limit Itô diffusion and write it both in Itô and Stratonovich form.
3. Problem 11.1 page 254 Øksendal.
4. **Variance reduction:**

Consider the sde

$$dX_t = \mu(X_t)dt + \sigma(X_t)d\beta_t,$$

and  $u(t, x) = \mathbb{E}^x[f(X_t)]$  with  $\mu, \sigma, f$  smooth and bounded, and  $x, \beta$  scalar valued. Assume that we can solve the SDE, but solving the pde problem for  $u$  is expensive (computing  $u$  may in particular be expensive in cases where the dimension of  $x$  is large). Define the Monte-Carlo estimate for  $u$ :

$$u(T, x) \approx \hat{u}^N = \frac{1}{N} \sum_{i=1}^N f(X_T^{(i)}),$$

with  $X_t^{(i)}$  independent realizations for  $X_t$  (assume the sde can be solved weakly “exactly”).

a) Show that  $\hat{u}^N$  is an unbiased estimate and that

$$\text{Var}(\hat{u}^N) = \frac{1}{N} \mathbb{E}^x \left[ \int_0^T \sigma^2(X_s) u_x^2(T-s, X_s) ds \right].$$

b) Consider

$$dY_t = (\mu(Y_t) + v(t, Y_t))dt + \sigma(Y_t)d\beta_t, \quad (1)$$

and assume uniform ellipticity:  $\sigma(x) > \Delta > 0$ . Denote the stochastic exponentials:

$$M_t^\pm = \mathcal{E} \left( \pm \int_0^t v(s, Y_s) \sigma(Y_s) d\beta_s \right),$$

(assume that  $v$  satisfies appropriate integrability conditions) and show that ( $\beta$  Brownian Motion under  $\mathbb{E}$ ):

$$u(t, x) = \mathbb{E}^x[f(X_t)] = \mathbb{E}^x[f(Y_t)M_T^-].$$

c) Introduce the estimator (given  $N$  independent realizations of the paths  $Y$ )

$$u(T, x) \approx \tilde{u}^N = \frac{1}{N} \sum_{i=1}^N f(Y_T^{(i)}) M_T^{-, (i)}.$$

d) Assume that  $f > \Delta > 0$  and show that  $v$  satisfying

$$\partial_x u(t, x) \sigma(x) = v(t, x) u(t, x) / \sigma(x), \quad (2)$$

is well defined. Show that with this choice of  $v$  in Eq. (1):  $\tilde{u}^1 = u(T, x)$ .

Hint: derive an expression for the variance of  $f(Y_T)M_T^-$  using Itô's rule and isometry, you may alternatively be able to show  $\tilde{u}^1 = u(T, x)$  directly, again using Itô calculus in view of the form of  $v$ .

Note that to find this “optimal control” or “importance sampler”  $v$  we need the function  $u$  which is the unknown. In practice we may use an estimate for  $u$  and an iteration in the Monte-Carlo estimation to improve this a priori estimate. (This process may in fact involve considering an augmented Markov process that allows for joint estimation of  $u, \partial_x u$ ).

d) Consider now  $v$  in Eq. (1) as a control. Set up an optimal control problem such that the optimal control is given by Eq. (2).

Hint: consider

$$V(t, x) = \min_v \mathbb{E}^{t,x} \left[ \int_t^T \sigma^2(Y_s) v_s^2 / 2 ds - \log(f(Y_T)) \right].$$

5. a) By considering the null-space of the generator for Brownian motion show that this process cannot have an invariant distribution.  
 b) For the  $n$ -dimensional Ornstein Uhlenbeck process show that the null-space is spanned by the constant functions and the null-space of the adjoint by the invariant distribution.  
 c) Set up a (pde) problem that (when solved) gives the probability that an OU process  $Y_t$  starting at  $y$ , ( $|y| < R$ ) satisfies  $\max_{0 < t < T} |Y_t| < R$ .
6. Problem 8.17 page 102 Karatzas & Schreve.
7. Problem 8.18 page 102 Karatzas & Schreve.
8. Consider Brownian Motion starting at  $x \in (0, 1)$  and reflected at  $0, 1$ .  
 a) Use the associated spectral representation (as in class) to show that the 1d Laplacian satisfies the Fredholm Alternative with respect to square integrable functions on  $[0, 1]$  satisfying homogeneous Neumann boundary conditions.  
 b) **The Resolvent & Shifting spectrum:**  
 Show that  $\Delta/2 - \lambda I$  ( $I$  the identity operator) is invertible w.r.t. functions square integrable on  $[0, 1]$  satisfying homogeneous Neumann boundary conditions, but not necessarily a solvability condition as in a).

Show that for such a function  $f$

$$((\Delta/2 - \lambda I)^{-1} f) = (-R_\lambda f)(x) = - \int_0^\infty \mathbb{E}^{0,x}[f(X_t)e^{-\lambda t} dt]$$

for appropriate  $X_t$  and comment on how we can interpret this in terms of a killed diffusion (which?) (here  $R_\lambda$  called resolvent). Express also  $(R_\lambda f)(x)$  in terms of the spectral decomposition in a).