Phase transitions in a control problem

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We consider the following controlled system of SDE

$$dX_t = \gamma dW_t + d\Lambda_t,$$

$$dY_t = X_t dW_t + d||\Lambda||_t - \beta dt,$$

where $\beta > 0$ and $\gamma \neq 0$ are constants, *W* is a standard Brownian motion and $||\Lambda||$ is the total variation of our control Λ that we require to be an adapted process of finite variation. The problem, motivated by a financial practice of hedging under transaction costs, is to minimize

$$\limsup_{T \to \infty} \frac{1}{T} E[Y_T^2] = \limsup_{T \to \infty} \frac{1}{T} E[\int_0^T (X_t^2 - 2\beta Y_t) dt + 2 \int_0^T Y_t d||\Lambda||_t].$$

This is a 2 dimensional stochastic control problem which is degenerate (both the Brownian motion W and the control Λ are only one dimensional). The optimal control is, by a formal dynamic programming principle, expected to be a singular one which keeps (X, Y) inside a region; however we have not yet had satisfactory results both from theoretical and practical points of view for this original problem. In this talk, we focus on a restricted class of control

$$d\Lambda = -\operatorname{sgn}(X_t)\gamma^2 c(X_t)dt + dL_t - dR_t,$$

where *c* is a nonnegative continuous even function on an interval [-b, b], b > 0 and *L* and *R* are nondecreasing processes with

$$dL_t = \mathbf{1}_{\{X_t = -b\}} dL_t, \quad dR_t = \mathbf{1}_{\{X_t = b\}} dR_t \tag{1}$$

which keep *X* stay in [-b, b]. Now our control is (b, c). The idea of the control is to push *X* towards 0. The absolutely continuous part of Λ determined by *c* pushes *X* regularly towards 0. The other parts, that turn out to be singularly continuous, are active only when *X* reaches the boundary of [-b, b] and push *X* to prevent it from going out of the interval. Such a control exists; in fact, there exists a pathwise unique strong solution (*X*, *L*, *R*) of a Skorokhod SDE

$$dX_t = \gamma dW_t - \operatorname{sgn}(X_t)\gamma^2 c(X_t)dt + dL_t - dR_t$$

on [-b, b] when $x \mapsto -\text{sgn}(x)c(x)$ is one-sided Lipschitz. The control Λ is then well-defined by (1). The optimal control in this restricted class is probably

suboptimal for the original problem; however it has a certain advantage in its easy implementation. Also this type of control strategies has appeared in a related context of optimal hedging. Now, we are no more sure about the validity of the dynamic programming principle due to the constraint that the control Λ can only refer to the spot value of X. Although we can formally derive an HJB type equation, it is far from a standard form and difficult to solve. Here we present our results based on a probabilistic approach.

Theorem:

1. As $T \to \infty$,

$$\frac{1}{\sqrt{T}}(Y_T - \delta^{b,c}T) \to \mathcal{N}(0, Q^{b,c})$$

in law, where

$$\delta^{b,c} = \frac{\gamma^2}{a} - \beta, \ a = 2 \int_0^b g(x) dx, \ g(x) = \exp\left\{-2 \int_0^x c(y) dy\right\}$$

and

$$Q^{b,c} = \frac{2}{a} \int_0^b (x - \gamma h(x))^2 g(x) dx, \quad h(x) = \frac{2}{g(x)} \int_0^x \left(c(y) - \frac{1}{a} \right) g(y) dy$$

2.

$$\lim_{T\to\infty}\frac{1}{T}E[(Y_T-\delta^{b,c}T)^2]=Q^{b,c}.$$

3.

$$\inf_{\delta^{b,c}=0} Q^{b,c} = \gamma^2 \eta\left(\frac{\gamma}{\beta}\right),$$

where

$$\eta(x) = \begin{cases} 0 & \text{if } -2 < x \le 1, \\ \frac{4}{3} \frac{(x+2)^2(x-1)}{x^3(4-x)} & \text{if } 1 < x < 2, \\ \frac{1}{12} (x+2)^2 & \text{if } |x| \ge 2. \end{cases}$$

The proof of the convergences is not difficult. The mathematically challenging part is the minimization of $Q^{b,c}$. We can give an explicit sequence of controls (b_n, c_n) with $\delta^{b_n,c_n} = 0$ such that Q^{b_n,c_n} converges to the infimum. In fact, $b_n = \gamma^2/2\beta$ and $c_n = 0$ when $|\gamma| \ge 2\beta$. When $|\gamma| < 2\beta$ on the other hand, $b_n \to \infty$ as $n \to \infty$ and the pointwise limit of $c_n(x)$ is given by

$$c_{\infty}(x) = \frac{\gamma + 2\beta}{2(\gamma - \beta l)} \frac{1}{\gamma + |x|} \mathbf{1}_{\{|x| \ge l\}}, \ l = \frac{2(\gamma - \beta)_+}{4\beta - \gamma}$$

The key for the minimization is to show

$$\inf_{y\in\mathcal{Y}_a}\int_0^1 \left(y(u)+\gamma+\frac{2\gamma}{a}(u-1)y'(u)\right)^2\mathrm{d} u=\gamma^2\eta(a/\gamma),$$

where \mathcal{Y}_a is the set of the convex functions on [0, 1] with y(0) = 0 and y'(0) = a/2.