LARGE DEVIATIONS FOR SOME FAST STOCHASTIC VOLATILITY MODELS BY VISCOSITY METHODS

Daria Ghilli University of Padua Joint work with Martino Bardi and Annalisa Cesaroni

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Large deviations for stochastic volatility models

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- 1 Stochastic volatility models
- 2 Large deviations by viscosity methods
- 3 As applications, asymptotic estimates for European out-of-the-money option prices near maturity and asymptotic formula for implied volatility.
- 4 Extension to the non-compact case (i.e. when the coefficients of the stochastic system are not periodic), work in progress.

STOCHASTIC SYSTEM WITH FAST OSCILLATING RANDOM PARAMETER

We consider a stochastic system in \mathbb{R}^n with random coefficients, in particular with coefficients dependent on random parameter Y_t .

$$dX_t = \phi(X_t, Y_t)dt + \sqrt{2}\sigma(X_t, Y_t)dW_t, \quad X_0 = x_0 \in \mathbb{R}^n.$$

Assumption: we model this new parameter as a markov process evolving on a faster time scale $\tau = \frac{t}{\delta}$:

$$dY_t = \frac{1}{\delta}b(Y_t)dt + \sqrt{\frac{2}{\delta}}\tau(Y_t)dW_t, \quad Y_0 = y_0 \in \mathbb{R}^m.$$

Notation: X_t are the slow components of the system, and Y_t are the fast components.

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Tyical Assumptions: The fast variable are constrained in a compact set, say: the coefficients of the processes are \mathbb{Z}^m -periodic with respect to the variable y.

More generally: Y_t is a recurrent process.

In particular we can generalize the results presented under the hypothesis that Y_t is ergodic, this means that Y forgets the initial condition for large time (i.e. as $\delta \to 0$) and its distribution becomes stationary.

For technical simplicity from now on we assume the condition of periodicity on the coefficients.

This condition can be relaxed to ergodicity and will be treated in an article in preparation.

Further assumption: the diffusion matrix τ is non-degenerate.

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MOTIVATION:: ANALYSIS OF FINANCIAL MODELS WITH STOCHASTIC VOLATILITY

Black-Scholes model: the evolution of the price of a stock S is described by

 $dlog S_t = \gamma dt + \sigma dW_t$, t=time, W_t = Wiener proc.,

and the classical Black-Scholes formula for option pricing is derived assuming parameters are constants.

In reality the parameters of such models are not constants. In particular, the volatility σ , a measure for variation of price over time, is not constant but exhibits random behaviour.

Therefore it has been modeled as a positive function $\sigma = \sigma(Y_t)$ of a stochastic process Y_t with

- 1 negative correlation (prices go up when volatility goes down)
- 2 mean reversion (the time it takes for agents to adjust their thresholds to current market conditions)

Refs.: Hull-White 87, Heston 93, Fouque-Papanicolaou-Sircar 2000,...



Multiscale stochastic volatility

It is argued in the book **Fouque, Papanicolaou, Sircar:** Derivatives in financial markets with stochastic volatility, 2000, that Y_t also evolves on a faster time scale than the stock prices, modelling better the typical bursty behavior of volatility, see previous picture.

For this reason we put ourselves into the framework of multiple time scale systems and sigular perturbation and we model Y_t with the fast stochastic process for $\delta > 0$

$$dY_t = \frac{1}{\delta}b(Y_t)dt + \sqrt{\frac{2}{\delta}}\tau(Y_t)dW_t \quad Y_0 = y_0 \in \mathbb{R}^m.$$

Passing to the limit as $\delta \to 0$ is a classical singular perturbation problem, its solution leads to the elimination of the state variable Y_t and to the definition of an averaged system defined in \mathbb{R}^n only. There is a large literature on the subject (**Bensoussan, Kushner, Hasminskii, Pardoux, Borkar, Galtsgory, Alvarex, Bardi...**)

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We study the small time behaviour of the system, so we rescale time as

$$t \to \varepsilon t$$
.

We study the asymptotics when both parameters go to 0 and we expect different limit behaviors depending on the rate ε/δ . Therefore we put

 $\delta = \varepsilon^{\alpha}$, with $\alpha > 1$.

We consider the limit of the system for $\varepsilon \to 0$

$$\begin{cases} dX_t^{\varepsilon} = \varepsilon \phi(X_t^{\varepsilon}, Y_t^{\varepsilon}) dt + \sqrt{2\varepsilon} \sigma(X_t^{\varepsilon}, Y_t^{\varepsilon}) dW_t, & X_0^{\varepsilon} = x_0 \in \mathbb{R}^n \\ dY_t^{\varepsilon} = \frac{1}{\varepsilon^{\alpha - 1}} b(Y_t^{\varepsilon}) dt + \sqrt{\frac{2}{\varepsilon^{\alpha - 1}}} \tau(Y_t^{\varepsilon}) dW_t, & Y_0^{\varepsilon} = y_0 \in \mathbb{R}^m. \end{cases}$$
(1)

Avellaneda and collaborators (2002, 2003) used the theory of large deviations to give asymptotic estimates for the Black-Scholes implied volatility of option prices near maturity (small time) in models with constant (local) volatility.

We carry on the same type of analysis in models with stochastic volatility. In this case finding explicit estimates happens to be more difficult and we need to assume condition of periodicity/ergodicity on the fast process.

Remark

In this model:

- ε : short maturity of the option
- $\delta = \varepsilon^{\alpha}$: rate of mean reversion of the volatility.

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1 J. Feng, J.-P. Fouque, R. Kumar (2012) studied large deviations for systems of the form that we defined for $\alpha = 2, 4$ in the one-dimensional case n = m = 1, assuming that Y_t is an Ornstein-Uhlenbeck process and the coefficients in the equation for X_t do not depend on X_t . The methods are based on the monograph by Feng and Kurtz, Large deviations for stochastic processes 2006.

2 Related works by **P. Dupuis, K. Spiliopoulos, K. Spiliopoulos** (2012, 2013) deal with different scaling and use different methods based on weak convergence

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Let $\{\mu_{\varepsilon}\}$ be a family of probability measures. A large deviation principle (LDP) characterizes the limiting behavior, as $\epsilon \to 0$, of $\{\mu_{\epsilon}\}$ in terms of a rate function through asymptotic upper and lower exponential bounds on the values that μ_{ϵ} assigns to measurable subsets of \mathbb{R}^{n} .

Roughly speaking, large deviation theory concerns itself with the exponential decline of the probability measures of certain kinds of extreme or tail events.

In the context of financial mathematics, large deviations theory arises in the computations of small maturity out-of-the-money option prices.

We prove a Large Deviation Principle (LDP) for the process X_t^{ε} (i.e. for probability measures generated by the laws of X_t^{ε}).

In other words we prove that then for every t > 0 and for any open set $B \subseteq \mathbb{R}^n$

$$P(X_t^{\varepsilon} \in B) = e^{-\inf_{x \in B} \frac{I(x;x_0,t)}{\varepsilon} + o(\frac{1}{\varepsilon})}, \text{ as } \varepsilon \to 0.$$

for some (good) rate function I, non-negative and continuous, which we will define in the next slides.

BRYC'S INVERSE VARADHAN LEMMA

Assume that for all t > 0

- 1 X_t^{ε} is exponentially tight.
- $2\,$ for every h bounded and continuous the limit

$$\lim_{\varepsilon \to 0} \varepsilon \log E\left[e^{\varepsilon^{-1}h(X_t^{\varepsilon})} \,|\, X_0 = x_0, Y_0 = y_0\right] := L_h(x_0, t)$$

exists finite.

Then X_t^{ε} satisfies a large deviation principle with good rate function

$$I(x, x_0, t) = \sup_{h \in BC(\mathbb{R}^n)} \{h(x) - L_h(x_0, t)\}.$$

Large deviations for stochastic volatility models

We define the following logarithmic payoff

$$v^{\varepsilon}(t, x_0, y_0) := \varepsilon \log E\left[e^{\varepsilon^{-1}h(X_t^{\varepsilon})} \mid X_0 = x_0, Y_0 = y_0\right], x_0 \in \mathbb{R}^n, y_0 \in \mathbb{R}^m, t \ge 0,$$

where h is a bounded continuous function defined on \mathbb{R}^n .

Then, to obtain the LDP we have:

- 1 prove that v^{ε} converges to some function v(t, x) and characterize v;
- 2 compute the rate function I in term of the limit of v^{ε} .

The associated HJB equation to v_{ε} is the following parabolic pde with quadratic nonlinearity in the gradient $(b, \tau \text{ computed in } y, \phi, \sigma \text{ in } (x, y))$.

$$\begin{split} v_t^{\varepsilon} &= |\sigma^T D_x v^{\varepsilon}|^2 + \varepsilon \left(\operatorname{tr}(\sigma \sigma^T D_{xx}^2 v^{\varepsilon}) + \phi \cdot D_x v^{\varepsilon} \right) + 2\varepsilon^{-\frac{\alpha}{2}} (\tau \sigma^T D_x v^{\varepsilon}) \cdot D_y v^{\varepsilon} + \\ &+ 2\varepsilon^{1-\frac{\alpha}{2}} \operatorname{tr}(\sigma \tau^T D_{xy}^2 v^{\varepsilon}) + \varepsilon^{1-\alpha} \left(b \cdot D_y v^{\varepsilon} + \operatorname{tr}(\tau \tau^T D_{yy}^2 v^{\varepsilon}) \right) + \varepsilon^{-\alpha} |\tau^T D_y v^{\varepsilon}|^2. \end{split}$$

Remark

This problem falls in the class of averaging/homogenization problems for nonlinear HJB type equations where the fast variable lives in a compact space

THEOREM-CONVERGENCE BY VISCOSITY METHODS

Let h be continuous and bounded. Then

$$v^{\varepsilon}(x,y,t) = \varepsilon \log E e^{\frac{h(X_t^{\varepsilon})}{\varepsilon}} \to v(x,t)$$

locally uniformly in \boldsymbol{y} where \boldsymbol{v} is the unique viscosity solution to the effective equation

$$\begin{cases} v_t - \bar{H}(x, Dv) = 0 & \text{in }]0, T[\times \mathbb{R}^n, \\ v(0, x) = h(x) & \text{in } \mathbb{R}^n. \end{cases}$$

where \overline{H} is the limit or effective Hamiltonian.

Remark

- v^{ε} is uniformly bounded in ε .
- To prove the convergence we use relaxed semilimits Barles-Parthame procedure and the techniques stem form Evans' perturbed test function method for homogenization and its extension to singular perturbations Alvarez, Bardi (2003) and regular perturbations of singular perturbation Alvarez, Bardi, Marchi (2007).

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We identify the limit or effective Hamiltonian, by solving three different cell problems depending on α . We point out three regimes depending on how fast the volatility oscillates relative to the horizon length:

 $\left\{ \begin{array}{ll} \alpha > 2 & {\rm supercritical \ case}, \\ \alpha = 2 & {\rm critical \ case}, \\ \alpha < 2 & {\rm subcritical \ case}. \end{array} \right.$

- 1 In all the cases the limit Hamiltonian \overline{H} is continuous on $\mathbb{R}^n \times \mathbb{R}^n$ and convex in the second variable.
- 2 In all the cases we provide some representation formulas for the limit Hamiltonian \bar{H} .

More interesting case: $\alpha = 2$.

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 \bar{H} is continuous, convex,

$$\inf_{y} |\sigma^{T}(\bar{x}, y)\bar{p}|^{2} \le \bar{H}(\bar{x}, \bar{p}) \le \sup_{y} |\sigma^{T}(\bar{x}, y)\bar{p}|^{2}.$$

More precisely

$$\bar{H}(\bar{x},\bar{p}) = \lim_{t \to \infty} \frac{1}{t} \log E\left[e^{\int_0^t |\sigma^T(\bar{x},Y_s)\bar{p}|^2 \, ds} \,|\, Y_0 = y\right],$$

where Y_t is the stochastic process defined by

$$dY_t = \left(b(Y_t) + 2\tau(Y_t)\sigma^T(\bar{x}, Y_t)\bar{p}\right)dt + \sqrt{2}\tau(Y_t)dW_t$$

Large deviations for stochastic volatility models

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BRYC'S INVERSE VARADHAN LEMMA

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$$\lim_{\varepsilon \to 0} \varepsilon \log E\left[e^{\varepsilon^{-1}h(X_t^{\varepsilon})} \mid X_0 = x_0, Y_0 = y_0\right] := L_h(x_0, t)$$

exists finite.

Then X_t^{ε} satisfies a large deviation principle with good rate function

$$I(x, x_0, t) = \sup_{h \in BC(\mathbb{R}^n)} \{h(x) - L_h(x_0, t)\}.$$

Large deviations for stochastic volatility models

Throughout the section we suppose that σ is uniformly non degenerate, that is, for some $\nu > 0$ and for all $x, p \in \mathbb{R}^n$

$$|\sigma^T(x,y)p|^2 > \nu |p|^2.$$
 (2)

Note that under the previous assumption, the effective Hamiltonian is coercive. Let \overline{L} be the effective Lagrangian associated to the effective Hamiltonian \overline{H} via convex duality, i.e. for $x \in \mathbb{R}^n$

$$\bar{L}(x,q) = \max_{p \in \mathbb{R}^n} \{ p \cdot q - \bar{H}(x,p) \}.$$

Note that $\overline{L}(x, \cdot)$ is a convex nonnegative function such that $\overline{L}(x, 0) = 0$ for all $x \in \mathbb{R}^n$, since $\overline{H}(x, \cdot)$ is convex nonnegative and $\overline{H}(x, 0) = 0$ for all $x \in \mathbb{R}$.

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THE RATE FUNCTION

Then the rate function is defined as follows

$$I(x; x_0, t) := \inf \left[\int_0^t \bar{L}(\xi(s), \dot{\xi}(s)) \, ds \big| \xi \in AC(0, t), \ \xi(0) = x_0, \xi(t) = x \right].$$

- I depends only on the volatility σ and on the fast process Y_t^{ε} ;
- I does not depend on the drift ϕ of the log-price X_t^{ε} and on the initial value y_0 of the process Y_t .
- I satisfies the following growth condition for some $\nu, C > 0$ and all $x, x_0 \in \mathbb{R}^n$

$$\frac{1}{4C}\frac{|x-x_0|^2}{t} \le I(x;x_0,t) \le \frac{1}{4\nu}\frac{|x-x_0|^2}{t};$$

• if σ does not depend on x, i.e. $\bar{H} = \bar{H}(p)$, the rate function is

$$I(x; x_0, t) = t\overline{L}\left(\frac{x - x_0}{t}\right).$$

THE RATE FUNCTION

• If $\alpha > 2$ and n = 1 and $\overline{H} = \overline{H}(p)$, then

$$I(x;x_0,t) = \frac{|x-x_0|^2}{4\bar{\sigma}^2 t}$$
(3)

where

$$\bar{\sigma} = \sqrt{\int_{\mathbb{T}^m} \sigma(y)^2 d\mu(y)}$$

and μ is the invariant measure of the process Y_t defined in the previous slides, i.e.

$$dY_t = b(Y_t)dt + \sqrt{2}\tau(Y_t)dW_t,$$

REMARK

We observe that the rate function defined in (3) is the same as the rate function for the Black-Scholes model with constant volatility $\bar{\sigma}$. In other words, in the ultra fast regime, to the leading order, it is the same as averaging first and then taking the short maturity limit.

APPLICATIONS-OUT-OF-THE-MONEY OPTION PRICING

Let S_t^ε be the asset price, evolving according to the following stochastic differential system

$$\begin{cases} dS_t^{\varepsilon} = \varepsilon \xi(S_t^{\varepsilon}, Y_t^{\varepsilon}) S_t^{\varepsilon} dt + \sqrt{2\varepsilon} \zeta(S_t^{\varepsilon}, Y_t^{\varepsilon}) S_t^{\varepsilon} dW_t & S_0^{\varepsilon} = S_0 \in \mathbb{R}_+ \\ dY_t^{\varepsilon} = \varepsilon^{1-\alpha} b(Y_t^{\varepsilon}) dt + \sqrt{2\varepsilon^{1-\alpha}} \tau(Y_t^{\varepsilon}) dW_t & Y_0^{\varepsilon} = y_0 \in \mathbb{R}^m, \end{cases}$$
(4)

where $\alpha > 1$, τ, b are \mathbb{Z}^m -periodic in y with τ non-degenerate and $\xi : \mathbb{R}_+ \times \mathbb{R}^m \to \mathbb{R}, \, \zeta : \mathbb{R}_+ \times \mathbb{R}^m \to \mathbf{M}^{1,r}$ are Lipschitz continuous bounded functions, periodic in y.

Observe that $S_t^{\varepsilon} > 0$ almost surely if $S_0 > 0$.

We consider out-of-the-money call option with strike price K and short maturity time $T = \epsilon t$, by taking

$$S_0 < K$$
 or $x_0 < \log K$.

Similarly, by considering out-of-the-money put options, one can obtain the same for $S_0 > K$.

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As an application of the Large Deviation Principle, we prove

COROLLARY

For fixed t > 0

$$\lim_{\varepsilon \to 0^+} \varepsilon \log E\left[\left(S_t^{\varepsilon} - K\right)^+\right] = -\inf_{y > \log K} I\left(y; x_0, t\right).$$

When $\zeta(s, y) = \zeta(y)$, the option price estimate reads

$$\lim_{\varepsilon \to 0^+} \varepsilon \log E\left[(S_t^{\varepsilon} - K)^+ \right] = -I\left(\log K; x_0, t\right)$$

Large deviations for stochastic volatility models

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We recall that given an observed European call option price for a contract with strike price K and expiration date T, the implied volatility σ is defined to be the value of the volatility parameter that must go into the Black-Scholes formula to match the observed price.

We consider out-of-the-money European call option, with strike price K, and we denote by $\sigma_{\varepsilon}(t, \log K, x_0)$ the implied volatility.

APPLICATIONS-AN ASYMPTOTIC FORMULA FOR IMPLIED VOLATILITY

As a further application, we prove

COROLLARY

$$\lim_{\varepsilon \to 0^+} \sigma_{\varepsilon}^2(t, \log K, x_0) = \frac{(\log K - x_0)^2}{2 \inf_{y > \log K} I(y; x_0, t)t}$$

Note that the infimum in the right-hand side, is always positive by the assumption on S_0 and by the growth of the rate function.

REMARK

When $\alpha > 2$, the implied volatility is $\bar{\sigma}$ that is

$$\bar{\sigma} = \sqrt{\int_{\mathbb{T}^m} \sigma^2(y) d\mu(y)}.$$

Large deviations for stochastic volatility models

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Thank you for the attention!

Large deviations for stochastic volatility models

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The cell problem and the effective Hamiltonian, $\alpha = 2$

Pluggin in the equation the formal asymptotic expansion

$$v^{\varepsilon}(t, x, y) = v^{0}(t, x) + \varepsilon w(t, x, y).$$

we obtain

$$v_t^0 - |\sigma^T D_x v^0|^2 - 2(\tau \sigma^T D_x v^0) \cdot D_y w - b \cdot D_y w - |\tau^T D_y w|^2 - \operatorname{tr}(\tau \tau^T D_{yy}^2 w) = O(\varepsilon).$$

Proposition

For any fixed (\bar{x}, \bar{p}) , there exists a unique $H(\bar{x}, \bar{p})$ for which the uniformly elliptic equation with quadratic nonlinearity in the gradient

 $\bar{H}(\bar{x},\bar{p}) - |\sigma^T \bar{p}|^2 - \left(2\tau \sigma^T \bar{p} + b\right) \cdot D_y w(y) - |\tau^T D_y w(y)|^2 - tr(\tau \tau^T D_{yy}^2 w(y)) = 0,$

has a periodic viscosity solution w.

The cell problem and the effective Hamiltonian, $\alpha = 2$

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PROPOSITION

For any fixed (\bar{x}, \bar{p}) , there exists a unique $\bar{H}(\bar{x}, \bar{p})$ for which the uniformly elliptic equation with quadratic nonlinearity in the gradient

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\overline{H} : First Representation formula

• \bar{H} can be represented through stochastic control as

$$\bar{H}(\bar{x},\bar{p}) = \lim_{\delta \to 0} \sup_{\beta(\cdot)} \delta E\left[\int_0^\infty \left(|\sigma(\bar{x},Z_t)^T \bar{p}|^2 - |\beta(t)|^2 \right) e^{-\delta t} dt \,|\, Z_0 = z \right]$$

and

$$\bar{H}(\bar{x},\bar{p}) = \lim_{t \to \infty} \sup_{\beta(\cdot)} \frac{1}{t} E\left[\int_0^t (|\sigma^T(\bar{x},Z_s)\bar{p}|^2 - |\beta(s)|^2) ds \,|\, Z_0 = z \right],$$

where $\beta(\cdot)$ is an admissible control process taking values in \mathbb{R}^r for the stochastic control system

$$dZ_t = \left(b(Z_t) + 2\tau(Z_t)\sigma^T(\bar{x}, Z_t)\bar{p} - 2\tau(Z_t)\beta(t)\right)dt + \sqrt{2}\tau(Z_t)dW_t; \quad (5)$$

• Moreover

$$\bar{H} = \int_{\mathbb{T}^m} \left(|\sigma(\bar{x}, z)^T \bar{p}|^2 - |\tau(z)^T D w(z)|^2 \right) d\mu(z),$$

where $w = w(\cdot; \bar{x}, \bar{p})$ is the smooth solution to

$$\begin{split} \bar{H}(\bar{x},\bar{p}) - \operatorname{tr}(\tau\tau^T D_{yy}^2 w) &- |\tau^T D_y w|^2 + \\ &- (2\tau\sigma^T \bar{p} + b) \cdot D_y w - |\sigma^T \bar{p}|^2 = 0 \quad \text{in } \mathbb{R}^m \end{split}$$

and $\mu = \mu(\cdot; \bar{x}, \bar{p})$ invariant probability measure on the torus \mathbb{T}^m of the process (5) with the feedback $\beta(z) = -\tau^T(z)Dw(z)$, i.e.

$$dZ_t = \left(b(Z_t) + 2\tau(Z_t)\sigma^T(\bar{x}, Z_t)\bar{p} + 2\tau(Z_t)\tau^T(Z_t)Dw(Z_t)\right)dt + \sqrt{2}\tau(Z_t)dW_t.$$

• Moreover

$$\bar{H}(\bar{x},\bar{p}) = \lim_{t \to \infty} \frac{1}{t} \log E \left[e^{\int_0^t |\sigma^T(\bar{x},Y_s)\bar{p}|^2 \, ds} \, | \, Y_0 = y \right],$$

where Y_t is the stochastic process defined by

$$dY_t = \left(b(Y_t) + 2\tau(Y_t)\sigma^T(\bar{x}, Y_t)\bar{p}\right)dt + \sqrt{2}\tau(Y_t)dW_t$$

Sketch of the proof:

Take $v = v(t, x; \bar{x}, \bar{p})$ a periodic solution of the *t*-cell problem and define the function $f(t, y) = e^{v(t,y)}$. Then *f* solves the following equation

$$\begin{cases} \frac{\partial f}{\partial t} - f |\sigma^T \bar{p}|^2 - (2\tau \sigma^T \bar{p} + b) \cdot Df - \operatorname{tr}(\tau \tau^T D^2 f) = 0 & \text{in } (0, \infty) \times \mathbb{R}^m\\ f(0, z) = 1 & \text{in } \mathbb{R}^m. \end{cases}$$

and we conclude using the Feynam-Kac formula.

• Moreover

$$\bar{H}(\bar{x},\bar{p}) = \lim_{t \to \infty} \frac{1}{t} \log E \left[e^{\int_0^t |\sigma^T(\bar{x},Y_s)\bar{p}|^2 \, ds} \, | \, Y_0 = y \right],$$

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Take $v = v(t, x; \bar{x}, \bar{p})$ a periodic solution of the *t*-cell problem and define the function $f(t, y) = e^{v(t, y)}$. Then f solves the following equation

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The cell problem and the effective Hamiltonian, $\alpha > 2$

Plugging the formal asymptotic expansion

$$v^{\varepsilon}(t, x, y) = v^{0}(t, x) + \varepsilon^{\alpha - 1}w(t, x, y)$$

in the equation we get

$$v_t^0 = |\sigma^T D_x v^0|^2 + b \cdot D_y w + \operatorname{tr}(\tau \tau^T D_{yy}^2 w) + O(\varepsilon).$$

PROPOSITION

For each (\bar{x}, \bar{p}) fixed, there exists a unique constant $\bar{H}(\bar{x}, \bar{p})$ such that the linear second order uniformly elliptic equation

$$\bar{H}(\bar{x},\bar{p}) - tr(\tau\tau(y)^T D_{yy}^2 w_{\delta}(y)) - b(y) \cdot D_y w_{\delta}(y) - |\sigma(\bar{x},y)^T \bar{p}|^2 = 0 \text{ in } \mathbb{R}^m,$$

has a periodic smooth solution.

 \bar{H} can be represented as

$$\bar{H} = \int_{\mathbb{T}^m} |\sigma(\bar{x}, y)^T \bar{p}|^2 \, d\mu(y),$$

where μ is the invariant probability measure on the torus \mathbb{T}^m of the stochastic process

$$dY_t = b(Y_t)dt + \sqrt{2}\tau(Y_t)dW_t,$$

that is, the periodic solution of

$$-\sum_{i,j}\frac{\partial^2}{\partial y_i\partial y_j}((\tau\tau^T)_{ij}(y))\mu + \sum_i\frac{\partial}{\partial y_i}(b_i(y))\mu = 0 \quad \text{in } \mathbb{R}^m,$$

with $\int_{\mathbb{T}^n} \mu(y) \, dy = 1.$

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When n = 1,

$$\bar{H}(\bar{x},\bar{p}) = (\bar{\sigma}\bar{p})^2$$

where

$$\bar{\sigma}(\bar{x}) = \sqrt{\int_{\mathbb{T}^m} \sigma^2(\bar{x}, y) d\mu(y)}$$

and μ is the invariant measure of the following process Y_t

$$dY_t = b(Y_t)dt + \sqrt{2}\tau(Y_t)dW_t,$$

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 SOC

The cell problem and the effective Hamiltonian, $\alpha < 2$

We plug in the equation the formal asympthotic expansion

$$v^{\varepsilon}(t, x, y) = v^{0}(t, x) + \varepsilon^{\frac{\alpha}{2}} w(t, x, y).$$

and we obtain

$$v_t^0 = |\sigma^T D_x v^0|^2 + 2(\tau \sigma^T D_x v^0) \cdot D_y w + |\tau^T D_y w|^2 + O(\varepsilon)$$

Proposition

For any fixed (\bar{x}, \bar{p}) , there exists a unique constant $\bar{H}(\bar{x}, \bar{p})$ such that the first order coercive equation

$$\bar{H}(\bar{x},\bar{p}) - |\tau^T(y)D_yw(y) + \sigma^T(\bar{x},y)\bar{p}|^2 = 0 \quad in \ \mathbb{R}^m$$

admits a (Lipschitz continuous) periodic viscosity solution w.

\overline{H} : Representation formulas

• \bar{H} satisfies

$$\bar{H}(\bar{x},\bar{p}) = \lim_{\delta \to 0} \sup_{\beta(\cdot)} \delta \int_0^{+\infty} \left(|\sigma(\bar{x},y(t))^T \bar{p}|^2 - |\beta(t)|^2 \right) e^{-\delta t} dt,$$

where $\beta(\cdot)$ varies over measurable functions taking values in \mathbb{R}^r , $y(\cdot)$ is the trajectory of the control system

$$\begin{cases} \dot{y}(t) = 2\tau(y(t))\sigma^T(\bar{x}, y(t))\bar{p} - 2\tau(y(t))\beta, \quad t > 0, \\ y(0) = y \end{cases}$$

and the limit is uniform with respect to the initial position y of the system;

• Moreover under the condition $\tau \sigma^T = 0$ of non-correlations among the components of the white noise acting on the slow and the fast variables in the system, we have

$$\bar{H}(x,p) = \max_{y \in \mathbb{R}^m} |\sigma^T(x,y)p|^2.$$

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Remarks on the comparison principle for \bar{H}

• $\alpha = 2$: we relate the regularity in x of \overline{H} with that of the pseudo-coercive Hamiltonian $|\sigma^T(x, y)p|^2$, i.e. we prove for $0 < \mu < 1$

$$\mu \bar{H}(x,\frac{p}{\mu}) - \bar{H}(z,q) \ge \frac{1}{\mu-1} \sup_{y \in \mathbb{R}^m} |\sigma^T(x,y)p - \sigma^T(z,y)q|^2.$$

Then, we follow the same argument for pseudo-coercive Hamiltonian as in Barles-Perthame (1990) for the stationary case and M.Kobylanski for the evolutionary case (PhD thesis).

α > 2: as for the critical case (easier thanks to the explicit formula or H).
α < 2: let H₀ = √H. Then

 $H_0(x,\lambda p) = |\lambda| H_0(x,p) \quad \forall \lambda \in \mathbb{R}$

and there exists C > 0 such that $|H_0(x, p)| \le C|p|$ and

 $|H_0(x,p) - H_0(z,p)| \le C(1+|p|)|x-z| \quad \forall x,z \in \mathbb{R}^n, p \in \mathbb{R}^n.$

Then we can use the comparison result for evolutive non-coercive first-order HJ equations by Cutri, Da Lio (2007).

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Remarks on the comparison principle for H

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WITHOUT PERIODICITY ASSUMPTION (WORK IN PROGRESS)

Main assumptions: ergodicity of the fast variables

• $\exists C > 0$ ad a compact set K s.t.

$$b(y) \cdot y + 2\operatorname{tr}(\tau \tau^T(y)) < -C|y|^2 \quad \forall y \notin K.$$

Main example: Ornstein-Uhlenbeck process

$$dY_t = (m - Y_t)dt + \tau dW_t.$$

• τ is non degenerate and $\sigma(x, \cdot)$ is bounded $\forall x$.

Recall that v^{ε} solves

$$\begin{split} v_t^{\varepsilon} &= |\sigma^T D_x v^{\varepsilon}|^2 + \varepsilon \left(\operatorname{tr}(\sigma \sigma^T D_{xx}^2 v^{\varepsilon}) + \phi \cdot D_x v^{\varepsilon} \right) + 2\operatorname{tr}(\sigma \tau^T D_{xy}^2 v^{\varepsilon}) + \\ &\quad + \frac{1}{\varepsilon} \left((b + 2(\tau \sigma^T D_x v^{\varepsilon})) \cdot D_y v^{\varepsilon} + \operatorname{tr}(\tau \tau^T D_{yy}^2 V^{\varepsilon}) \right) + \frac{1}{\varepsilon^2} |\tau^T D_y v^{\varepsilon}|^2. \end{split}$$

and v^{ε} is uniformly bounded in ε . We will use relaxed semilimits (Barles-Perthame procedure).

We want to prove that v^{ε} converges locally uniformly to the unique solution to

$$v_t - \bar{H}(x, Dv) = 0.$$

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For every x, p fixed, find the unique constant $\bar{H}(x, p)$ such that there is a solution w to

$$\begin{cases} -(b+2\tau\sigma^{T}p)\cdot Dw - \operatorname{tr}(\tau\tau^{T}D^{2}w) + |\tau^{T}Dw|^{2} + |\sigma^{t}p|^{2} = -\bar{H} \\ \exists C > 0 \text{ s.t. } w(y) \le C(1+|y|^{2}) \end{cases}$$
(6)

- The main assumption on b, τ implies that $K(1 + |y|^2)$ is a supersolution to the previous problem outside a compact set.
- Without growth assumptions, no uniquenes of \bar{H} (Ichiara 2011, Khaise-Sheu 2006).

We look at the ergodic approximation

$$\delta v_{\delta} - (b + 2\tau \sigma^t p) \cdot Dv_{\delta} - tr(\tau \tau^T D^2 v_{\delta}) - |\tau^T Dv_{\delta}|^2 - |\sigma^t p|^2 = 0,$$

and prove that $\delta v_{\delta} \to \bar{H}, v_{\delta} - v_{\delta}(0) \to w.$

We consider approximate ergodic problems in balls with singular boudary conditions:

for every R >> 1, for every x, p fixed, find the unique constant $\overline{H}_R(x, p)$ such that there is a solution w_R to

$$\begin{cases} -(b+2\tau\sigma^T p) \cdot Dw_R - \operatorname{tr}(\tau\tau^T D^2 w_R) + |\tau^T Dw_R|^2 + |\sigma^t p|^2 = -\bar{H}_R \quad |y| < R\\ w_R \to -\infty \quad |y| \to R. \end{cases}$$

(Barles, Porretta, Tchamba, 2010).

Then \bar{H}_R is increasing to \bar{H} as $R \to +\infty$.

(7)

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Define

 $\underline{v}(t,x,y) = \inf(\liminf_{\varepsilon \to 0} v^{\varepsilon}(x_{\varepsilon},y_{\varepsilon},t_{\varepsilon}) \,|\, x_{\varepsilon} \to x, y_{\varepsilon} \to y, t_{\varepsilon} \to t).$

• for all t, x fixed, $\underline{v}(t, x, \cdot)$ is a (bdd) supersolution to

 $-|\tau^T D_y \underline{v}|^2 \ge 0.$

So, it does not depend on y (Liouville type theorem).

- If $\underline{v}(t,x) f(t,x)$ has a strict minimum at a point $(\overline{t},\overline{x})$, then $v^{\varepsilon}(t,x,y) f(t,x) \varepsilon w_R(y)$ has minima, for ε small, $x_{\varepsilon} \to \overline{x}, t_{\varepsilon} \to \overline{t}, y_{\varepsilon} \in B(0,R).$
- By the definition of viscosity solutions and the construction of w_R , we get that

$$\underline{v}_t - \bar{H}_R(x, D\underline{v}) \ge 0,$$

for every R >> 1.

$\alpha = 2$, Subsolution

Define

$$\bar{v}(t,x) = \sup(\limsup_{\varepsilon \to 0} v^{\varepsilon}(x_{\varepsilon}, y_{\varepsilon}, t_{\varepsilon}) \,|\, x_{\varepsilon} \to x, t_{\varepsilon} \to t, (y_{\varepsilon})_{\varepsilon} \text{ is bounded}).$$

Note that

$$\bar{v}(t,x) \geq \sup(\limsup_{\varepsilon \to 0} v^{\varepsilon}(x_{\varepsilon},y_{\varepsilon},t_{\varepsilon}) \,|\, x_{\varepsilon} \to x, y_{\varepsilon} \to y, t_{\varepsilon} \to t).$$

• We construct a supersolution \tilde{w} to the ergodic problem such that

•
$$\tilde{w} \to +\infty$$
 as $|y| \to +\infty$

• $-(b+2\tau\sigma^T p) \cdot D\tilde{w} - \operatorname{tr}(\tau\tau^T D^2\tilde{w}) - |\tau^T D\tilde{w}|^2 \to +\infty \text{ as } |y| \to +\infty.$

• If $\bar{v}(t,x) - f(t,x)$ has a strict maximum at a point (\bar{t},\bar{x}) , then $v^{\varepsilon}(t,x,y) - f(t,x) - \varepsilon \tilde{w}(y)$ has maxima, for ε small, $x_{\varepsilon} \to \bar{x}, t_{\varepsilon} \to \bar{t}$ and y_{ε} that is for all ε in a compact set independent of ε . Then, using the definition of viscosity solution,

$$\bar{v}_t - \bar{H}(x, D\bar{v}) \le 0.$$

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By construction

 $\overline{v}(t,x) \ge \underline{v}(x,t).$

We conclude using the comparison principle for the effective equation

$$v_t + \bar{H}(x, Dv) = 0$$

 that

$$v(x,t) = \sup(\limsup_{\varepsilon \to 0} v^{\varepsilon}(x_{\varepsilon}, y_{\varepsilon}, t_{\varepsilon}) | x_{\varepsilon} \to x, t_{\varepsilon} \to t, y_{\varepsilon} \text{ bounded})$$

= $\inf(\liminf_{\varepsilon \to 0} v^{\varepsilon}(x_{\varepsilon}, y_{\varepsilon}, t_{\varepsilon}) | x_{\varepsilon} \to x, t_{\varepsilon} \to t, y_{\varepsilon} \text{ bounded}).$ (8)

DQC

Let $\varepsilon > 0$ and consider u^{ε} solution to

$$\begin{cases} u_t^{\varepsilon} - b\left(\frac{x}{\varepsilon}\right) Du^{\varepsilon} - \varepsilon \mathrm{tr}(\tau \tau^T\left(\frac{x}{\varepsilon}\right) - |\tau\left(\frac{x}{\varepsilon}\right) D^2 u^{\varepsilon}|^2 - l\left(x, \frac{x}{\varepsilon}\right) = 0\\ u^{\varepsilon}(x, 0) = h(x). \end{cases}$$

If the main assumptions hold for b, τ (the underlying process to the fast variables is ergodic), then u^{ε} converges locally uniformly, as $\varepsilon \to 0$, to the solution u of

$$u_t - \bar{H}(x, Du) = 0.$$

Large deviations for stochastic volatility models