

# **Front Speeds of Non-Coercive Hamilton-Jacobi Equations in Multiscale Media**

**Jack Xin**

**Mathematics, UC Irvine.**

**Supported in part by NSF.**

Fronts or interfaces in multiscale and random flows appear in many scientific areas as solutions to PDEs or free boundary problems: premixed turbulent combustion, chemical reaction fronts in liquid, algal blooms (large blobs of micro-algae) in the ocean. Two prototype scalar models for fronts are:

I. Reaction-Diffusion-Advection-Equation (RD):

$$u_t = \Delta_x u + B(\omega, x, t) \cdot \nabla_x u + f(u), \quad x \in R^N, \quad (1)$$

$f(u)$  is nonlinear reaction function;  $B$  a space-time periodic or random flow field.

II. Hamilton-Jacobi Equation (HJ):

$$u_t + H(\omega, x, t, \nabla_x u) = \kappa \Delta u, \quad x \in R^N, \quad N \geq 1, \quad \kappa \geq 0. \quad (2)$$

$H$  is periodic or random in  $(x, t)$ , nonlinear in  $\nabla_x u$ .

### Examples of reactive nonlinearities:

1.  $f(u) = u(1 - u)$  (Kolmogorov-Petrovsky-Piskunov(KPP), Fisher, 1937)
2.  $f(u) = 0 \forall u \in [0, \theta] \cup \{1\}$ ,  $f(u) > 0 \forall u \in (\theta, 1)$ ,  $f(u)$  Lipschitz continuous, combustion nonlinearity with ignition temperature  $\theta$ .
3.  $f(u) = u(1 - u)(u - \mu)$ ,  $\mu \in (0, 1)$ : bistable nonlinearity.

### Examples of Hamiltonians:

1.  $H(x, t, p) = |p|^2/2 + V(x)$  (classical)
2.  $H(x, t, p) = |p|^2/2 + V(x) \cdot p$  (advective)
3.  $H(x, t, p) = |p| + V(x) \cdot p$  (relativistic/geometric)

**Consider Spatially Heterogeneous Hamilton-Jacobi Eq. (HJ) satisfying:**

- $H$  (Hamiltonian) is convex in  $p \equiv \nabla_x u$ , continuous in  $x$ ;
- $H$  is stationary in  $x$  (joint distribution of  $H(\omega, x_1, p), \dots, H(\omega, x_m, p)$  is invariant under translation in  $x$ );
- $H$  is ergodic in  $x$  (events invariant under translation have probability either zero or one).

Theory of HJ homogenization requires **coercivity**:

$$H(x, p) \rightarrow +\infty, \quad \text{as } p \rightarrow \infty \text{ uniformly in } x.$$

- Periodic Media: Lions-Papanicolaou-Varadhan (1988), Evans (92);
- Quasi-periodic media and degenerate 2nd order (viscous) HJs: (Lions-Souganidis, Ishii, and references therein).
- Stationary ergodic media:
  - Inviscid HJs: Souganidis (99), Rezakhanlou-Tarver (00);
  - Viscous HJs: Lions-Souganidis (05), Kosygina-Rezakhanlou-Varadhan (06), Schwab (08).

## Non-coercive HJs:

Example 1:  $H(x, p, \omega) = c_n |p|^n + V(x, \omega)$ ,  $n = 2$  (classical mechanics),  $n = 1$  (relativistic mechanics), potential  $V$  is an Ornstein-Uhlenbeck (OU) process, a Gaussian process with mean zero and covariance

$$\mathbf{E}[V(x)V(y)] = \frac{1}{2} \exp(-|x - y|).$$

The OU process is unbounded in  $x$ ,  $H$  is **non-coercive**.

Example 2 (advective Hamiltonian):  $H(x, p, \omega) = c_n |p|^n + b(x, \omega) \cdot p$ ,  $n = 2$  (corres. to KPP),  $n = 1$  (G-equation, approximating ignition/bistable reaction fronts on hyperbolic time scale).

If  $n = 1$ , then for  $b$  mean-zero periodic function with amplitude above  $c_n$ ,  $H$  is **non-coercive**.

### Inviscid G-equation:

$$G_t + B(x) \cdot \nabla G = s_l |\nabla G|,$$

$s_l > 0$  (constant, laminar speed) as a small diffusion fast reaction limit of ignition/bistable reaction:

$$v_t + B(x) \cdot \nabla v = \epsilon \Delta v + \epsilon^{-1} f(v).$$

Substituting:

$$v \sim U(u(x, t)/\epsilon),$$

then equation at order  $O(\epsilon^{-1})$  is:

$$(u_t + B(x) \cdot \nabla u)U' = |\nabla u|^2 U'' + f(U).$$

By traveling front identity and scaling:

$$U'' - cU' + f(U) = 0, \quad d^2 U'' - dcU' + f(U) = 0, \quad \forall d > 0,$$

we have:

$$u_t + B(x) \cdot \nabla u = c|\nabla u|.$$

Over longer (diffusive) time scale  $O(\epsilon^{-2})$ , one adds curvature correction to G-equation, giving viscous G-equation.

## Capture Front Asymptotically in Multiscale Media and Homogenization:

Scaling transform:  $(x, t) \rightarrow \frac{(x, t)}{\epsilon}$ ,  $u \rightarrow \epsilon u \equiv u^\epsilon$ . In case of homogeneous media, front solutions are invariant.

HJ random fronts may be studied in terms of  $u^\epsilon$  in the limit  $\epsilon \downarrow 0$ , where  $u^\epsilon$  satisfies:

$$u_t^\epsilon + H\left(\omega, \frac{x}{\epsilon}, \nabla_x u^\epsilon\right) = 0, \quad (3)$$

with planar initial data:  $u^\epsilon(x, 0) = p \cdot x$ .

The study of convergence of  $u^\epsilon$  to a limiting function  $\bar{u}(x, t)$  and the resulting limiting eqn is **Homogenization**.

Suppose  $\bar{u}$  exists, then observing at  $(x, t) = (0, 1)$ , we see:

$$\epsilon u\left(0, \frac{1}{\epsilon}\right) \rightarrow \bar{u}(0, 1) \equiv -\bar{H}(p),$$

$p$  dependent constant or HJ front speed ( $T = \frac{1}{\epsilon}$ ):

$$u(0, T) \sim -\bar{H}(p) T.$$

**Lax-Oleinik-Hopf formula:**

$$u^\epsilon(x, t) = \inf_{y \in \mathbb{R}^N} \left( u^\epsilon(y, 0) + \inf_{\xi} \int_0^t L(\omega, \xi(s)/\epsilon, \xi'(s)) ds \right), \quad (4)$$

$\xi \in W^{1,\infty}((0, t); \mathbb{R}^N)$ ,  $\xi(0) = y$ ,  $\xi(t) = x$ , a Lipschitz path connecting  $y$  to  $x$ .

$L$  is Lagrangian given by Legendre transform:

$$L(\omega, x, q) = \sup_{p \in \mathbb{R}^N} (q \cdot p - H(\omega, x, p)). \quad (5)$$

$L$  is a convex function in  $q$ .

At  $x = 0$ , the integral equals:

$$\int_0^t L(\omega, \xi(s)/\epsilon, -\xi'(s)) ds, \quad (6)$$

where  $\xi(0) = 0$ ,  $\xi(t) = y$ . Letting  $n = 1/\epsilon$ , we study the limit  $S_n(y, t)/n$ , where:

$$S_n(y, t) = \inf_{\xi \in W^{1,\infty}([0, nt]; \mathbb{R}^N): \xi(0)=0, \xi(nt)=ny} \int_0^{nt} L(\omega, \xi(s), -\xi'(s)) ds. \quad (7)$$



## Subadditive Ergodic Theorem and Convergence of $S_n/n$ to a Nonrandom Number.

Define:

$$S_{m,n}(y, t) = \inf_{\xi(mt)=my; \xi(nt)=ny} \int_{mt}^{nt} L(\omega, \xi(s), -\xi'(s)) ds. \quad (8)$$

with inf taken over all paths  $\xi$  connecting  $my$  to  $ny$  in time  $(n - m)t$ .

SET(Kingman): Suppose that  $S_{m,n}$  are random variables satisfying:

- (1)  $S_{0,0} = 0$ ,  $S_{m,n} \leq S_{m,k} + S_{k,n}$ , for  $m \leq k \leq n$ ;
- (2)  $\{S_{m,m+k}, m \geq 0, k \geq 0\}$  equals  $\{S_{m+1,m+k+1}, m \geq 0, k \geq 0\}$  in distribution;
- (3)  $E[S_{0,1}^+] < +\infty$ ;  $\alpha_n \equiv E[S_{0,n}] < \infty$ ;

then:

(1)

$$\alpha = \lim_{n \rightarrow \infty} \frac{\alpha_n}{n} = \inf_{n \geq 1} \frac{\alpha_n}{n} \in [-\infty, \infty);$$

(2)  $S_\infty = \lim_{n \rightarrow \infty} \frac{S_{0,n}}{n}$  exists with probability one;

(3) if  $E[S_\infty] = \alpha > -\infty$ ,  $\lim_{n \rightarrow \infty} E[|S_{0,n}/n - S_\infty|] = 0$ .

Condition (1) follows directly from infimum of the path integral.

Condition (2) follows from stationarity.

Condition (3) requires further knowledge of “unboundedness” of the random Lagrangian.

If  $H$  is uniformly bounded in  $x$  (for any  $p$ ), then  $L$  is uniformly bounded in  $x$ . All moment assumptions hold, SET implies the existence of the limit  $S_n/n$  with probability one. The limit is invariant under translations of the realization of  $L$  by vectors proportional to  $x$ , hence its value is almost surely nonrandom (ergodicity).

By absorbing  $t$  into  $n$ , we see that the limit is of the form:  $t\bar{L}(\frac{y}{t})$ , where  $\bar{L}$  (homogenized Lagrangian) is convex in  $y/t$  (velocity) due to convexity of  $L$  in  $\xi'$  (velocity) and subadditivity of  $S_n$ .

The homogenization limit of HJ solution from planar data at  $x = 0$  is:

$$\bar{u}(0, t) = \inf_y (p \cdot y + t\bar{L}(y/t)) = t \inf_{y'} (p \cdot y' + \bar{L}(y')) \equiv -t\bar{H}(p). \quad (9)$$

This further implies that in general the homogenized eqn is  $\bar{u}_t + \bar{H}(\nabla_x \bar{u}) = 0$ .

Uniform boundedness is however not sharp for HJ homogenization, and not true for Gaussian processes. Unboundedness is an indication of **extreme behavior of a stochastic system**.

Example I:  $H = |p|^2/2 + V(x, \omega)$ , suppose that  $V$  is uniformly bounded from above by  $V_0$  (but no uniform lower bound), and that  $E[|V(x)|] < \infty$ . Then homogenization still holds.

Lagrangian  $L(x, q) = \frac{|q|^2}{2} - V(x, \omega)$ . For a linear path  $\xi(s) = s \frac{y}{t}$ , we obtain

$$S_{0,n} \leq \frac{n |y|^2 t}{2} - \int_0^{nt} V\left(\frac{s y}{t}\right) ds.$$

Integrability of  $V(x)$  implies that

$$\mathbf{E}[|S_{0,n}(t, y)|] < +\infty.$$

On the other hand, since the kinetic part of Lagrangian integral is positive, upper bound on  $V$  implies that

$$E[S_{0,n}(t, y)] \geq -n t V_0.$$

The family  $S_{m,n}$  satisfies the assumptions of subadditive ergodic theorem, implying existence of the finite limit of  $\frac{S_{0,n}}{n}$ .

## Unbounded Potentials and Running Maxima

Let  $V$  be an Ornstein-Uhlenbeck process ( $N = 1$ ), the running maximum

$V^*(y) = \sup_{u \in [0, y]} V(u)$  obeys:

### Theorem 1 (Cramér)

$$\text{Prob} \left[ \frac{V^*(y) - b_y}{a_y} \leq x \right] \rightarrow \exp(-e^{-x})$$

as  $y \rightarrow \infty$ , where

$$a_y^{-1} \sim b_y \sim (2 \log y)^{\frac{1}{2}}$$

with  $\sim$  denoting asymptotic equivalence.

Similar results hold in  $N \geq 2$  (Leadbetter et al).

The theorem says that the renormalized random variables  $V^*(y)$  converge in distribution to the double exponential distribution. It follows that

$$\frac{V^*(y)}{(2 \log y)^{\frac{1}{2}}} \rightarrow 1, \quad \text{as } y \rightarrow \infty,$$

in probability.

**Scaling  $\sqrt{2 \log n}$  for running max of i.i.d. sequence  $X_n$ :**

$$\begin{aligned} \text{Prob}\left(\max_{1 \leq i \leq n} \{X_1, \dots, X_n\} \leq x\right) &= \text{Prob}(X_1 \leq x, X_2 \leq x, \dots, X_n \leq x) \\ &= [\text{Prob}(X_1 \leq x)]^n \end{aligned}$$

which converges to zero as  $n \rightarrow \infty$  for any finite  $x$ . Almost surely the running max is unbounded!

If  $X_1$  is unit Gaussian,  $x = \sqrt{2 \log n}$ ,

$$\begin{aligned} [\text{Prob}(X_1 \leq x)]^n &= \left(1 - (2\pi)^{-1/2} \int_{\sqrt{2 \log n}}^{\infty} \exp\{-y^2/2\} dy\right)^n \\ &\approx (1 - \text{const.}/n)^n, \end{aligned}$$

converging to a positive finite number, suggesting running max scales as  $O(\sqrt{2 \log n})$ .

## Unboundedness of $V$ from Above and Breakdown of Homogenization

$N = 1$ , minimum action integral is:

$$S_n = \inf_{u(\tau)} \int_0^{nt} \left[ \frac{1}{2} \left( \frac{du}{d\tau} \right)^2 - V(u(\tau)) \right] d\tau,$$

over  $C^1$  functions  $u(\tau)$ :  $u(0) = 0$  and  $u(nt) = nx$ . The minimizing function  $u$  is monotone increasing:  $\frac{du}{d\tau} \neq 0$  for all  $\tau$ .

Rewrite  $S_n$  in inverse function  $\tau(u)$ :

$$S_n = \inf_{\tau(u)} \int_0^{nx} \left[ \frac{1}{2} \left( \frac{d\tau}{du} \right)^{-1} - V(u) \frac{d\tau}{du} \right] du,$$

over all  $C^1$  functions  $\tau(u)$ :  $\tau(0) = 0$  and  $\tau(nx) = nt$ .

Conservation of energy gives:

$$\frac{d\tau}{du} = \frac{1}{\sqrt{2[E_n - V(u)]}},$$

$E_n$ : the total energy,  $E_n > \max_{u \in [0, nx]} V(u)$ .

It follows that

$$S_n = \int_0^{nx} \left[ \sqrt{\frac{E_n - V(u)}{2}} - \frac{V(u)}{\sqrt{2(E_n - V(u))}} \right] du, \quad (10)$$

and

$$\int_0^{nx} \frac{1}{\sqrt{2(E_n - V(u))}} du = nt. \quad (11)$$

If  $E_n > \sup_{u \in [0, nx]} V(u) + \frac{x^2}{2t^2}$ , then  $\int_0^{nx} \frac{1}{\sqrt{2(E_n - V(u))}} du < nt$ , contradicting (11).

Hence  $E_n \leq \sup_{u \in [0, nx]} V(u) + \frac{x^2}{2t^2}$ , and

$$E_n \sim \sup_{u \in [0, nx]} V(u), \quad n \gg 1.$$

Rewrite from (10):

$$S_n = \int_0^{nx} \left[ \sqrt{\frac{E_n - V(u)}{2}} + \frac{E_n - V(u)}{\sqrt{2(E_n - V(u))}} - \frac{E_n}{\sqrt{2(E_n - V(u))}} \right] du. \quad (12)$$

First two terms of integrand in (12) are identical. Jensen's inequality (concavity of root function):

$$\begin{aligned} \frac{1}{nx} \int_0^{nx} \sqrt{2(E_n - V(u))} du &\leq \sqrt{\frac{1}{nx} \int_0^{nx} 2(E_n - V(u)) du} \\ &= \sqrt{2E_n - 2\frac{1}{nx} \int_0^{nx} V(u) du}, \end{aligned}$$

where almost surely as  $n \rightarrow \infty$ ,  $\frac{1}{nx} \int_0^{nx} V(u) du \rightarrow 0$ , by ergodicity of the Ornstein-Uhlenbeck process.

The integral of the first two terms is bounded by  $3\sqrt{E_n}nx$  for large  $n$ . The integral of the third term is  $-E_n nt$ . So for  $n$  large:

$$-tE_n \leq \frac{1}{n} S_n \leq 3x\sqrt{E_n} - tE_n,$$

or:

$$\frac{S_n}{n} \sim -tE_n \sim -t \sup_{u \in [0, nx]} V(u) = -tV^*(nx) \rightarrow -\infty.$$



Standard homogenization limit fails. Instead, the modified limit holds in probability:

$$\frac{S_n}{n(2 \log n)^{1/2}} \rightarrow -t.$$

Divergence of homogenization means that for affine data, the growth rate of HJ solutions in time is faster than linear, or front acceleration. Similar divergence occurs in  $N \geq 2$ .

For classical mechanics Hamiltonian, stochastic homogenization holds if and only if the potential is uniformly bounded from above.

**For advective Hamiltonian, upper bound may or may not be the criterion.**

Example IIa (H in Gradient Flows):  $H(x, p) = \frac{p^2}{2} + p \cdot b(x)$ ,  $b = \nabla U(x)$ , where  $U(x)$  is a scalar random vector field whose realizations are of class  $\mathbf{C}^2$ ,  $p, x \in \mathbb{R}^N$ . Such fields  $b$  include Gaussian random fields with appropriate covariance.

Corresponding Lagrangian is

$$L(x, q) = \frac{1}{2}|q - b(x)|^2 = \frac{|q|^2}{2} - q \cdot b(x) + \frac{|b(x)|^2}{2}. \quad (13)$$

For a path  $\xi(s)$ ,  $0 \leq s \leq nt$ , such that  $\xi(0) = 0$  and  $\xi(nt) = nx$ , the contribution from the second term to Lagrangian integral equals

$$\int_0^{nt} b(\xi(s)) \cdot \xi'(s) ds = U(nx)$$

Such a path independent term is called a null Lagrangian. Thus Lagrangian (13) leads to the same Euler-Lagrange equations of motion for the minimizing path as the Lagrangian of the potential system:

$$L_1(x, q) = \frac{|q|^2}{2} - V_1(x),$$

where  $V_1(x) = -\frac{1}{2}|b|^2(x) \leq 0$ . By result of Example I, **homogenization holds** for such Hamiltonian of advection type, **even though flow field is unbounded !**

Example IIb (H in shear flows):  $b(x) = (V_2(x'), 0)$ ,  $x' = (x_2, \dots, x_N)$ ,  $0 \in R^{N-1}$ .

HJ Eqn is:

$$u_t + b(x) \cdot \nabla_x u + |\nabla_x u|^2/2 = 0. \quad (14)$$

Consider a front moving in the  $x_1$  direction in the form  $u(x, t) = x_1 - \frac{1}{2}t + w(x', t)$ . Then  $w$  satisfies HJ eqn:

$$w_t + |\nabla_{x'} w|^2/2 + V_2(x') = 0, \quad (15)$$

which is in the classical potential form.

If  $V_2$  obeys the assumptions in Example I and is unbounded from above,  $-w(x', t)/t$  diverges for large time  $t$ . Front speed acceleration appears due to dominance of running maxima of process  $V_2$ .

## Summary of HJ Fronts in Random Media:

- **Homogenization:** behavior of the system on large scales is described by an effective nonrandom Lagrangian (or Hamiltonian). **Disorder gets averaged, extreme nature of the random media is tamed, fronts move at asymptotically constant speeds.** An elementary (and linear) analog of this phenomenon in classical probability theory is the strong **law of large numbers**.
- **Domination by finite-volume maxima of the random potential: Homogenization breaks down.** On an arbitrarily large scale, behavior of the system is dictated by the maximum value of the disorder on that scale. **The extreme nature of the random media prevails.** Analogue in classical probability theory is the study of extrema of stochastic sequences and processes. For example, the maximum  $M_n$  of  $n$  independent unit normal random variables behaves asymptotically as  $\sqrt{2 \log n}$  as  $n \rightarrow \infty$ .

The convergence (homogenization) and divergence (dominance by running maxima) phenomena of stochastic H-J was noted earlier in KPP (Xin 2003).

The divergence of HJ homogenization and one-sided condition of random potentials for HJ homogenization are based on E-Wehr-Xin (2008).

## G-equation and Control Representation:

Suppose  $u$  is the viscosity solution of

$$u_t + V(x)\nabla u = s_l|\nabla u|, \quad u(x, 0) = u_0(x), \quad (16)$$

then:

$$u(x, t) = \sup u_0(y(t)) \quad (17)$$

where the supremum is over all  $y \in W^{1,\infty}([0, t]; \mathbb{R}^d)$  satisfying  $y(0) = x$  and the constraint

$$|y'(\tau) + V(y(\tau))| \leq s_l \quad (18)$$

for all  $\tau \in [0, t]$ .

Suppose  $w(x, t)$  satisfies:

$$w_t + V(x)\nabla w = s_l|\nabla w| + \kappa\Delta w, \quad w(x, 0) = w_0(x), \quad (19)$$

then  $w(x, t)$  has the representation:

$$w(x, t) = \sup \mathbb{E} [u_0(Y_t)] \quad (20)$$

where  $Y_t$  satisfies the stochastic equation

$$dY_\tau = (\alpha_\tau - V(Y_\tau)) d\tau + \sqrt{2\kappa} dW_\tau, \quad Y_0 = x. \quad (21)$$

The supremum in (20) is over all stochastic controls  $\alpha_\tau$  which are progressively measurable with respect to the Brownian filtration, and satisfy  $|\alpha_\tau| \leq s_l$ .

We shall use these control formulas to analyse front speeds. The following is based on joint work with J. Nolen (2009).

**G-eqn and Spatial Flow:**  $V = V(x)$

$$u_t + V(x) \cdot \nabla_x u = s_l |\nabla_x u|,$$

with affine initial data  $u(x, 0) = p \cdot x$ , where  $p$  a unit vector in  $R^n$ .

Control formula of solution is:

$$u(x, t) = \max_{\alpha} p \cdot y,$$

where  $y = y(t; x)$  is the solution to the ODE with continuous control  $\alpha$ ,  $|\alpha| \leq s_l$ :

$$y'(\tau) = -V(y(\tau)) + \alpha(\tau), \tag{22}$$

and initial data  $y(0) = x$ .

## Gradient flow

Let  $p = 1$ , by ODE comparison argument,  $\alpha = s_l$  is the optimal control facilitating largest velocity to the right. Solution is  $u(x, t) = y$ , where  $y = y(t; x)$  solves:

$$y'(\tau) = -V(y(\tau)) + s_l, \quad (23)$$

with initial data  $y(0) = x$ . Two regimes arise.

If  $\min -V(y) + s_l > 0$  for all  $y$ , closed form solution:

$$\int_x^y \frac{d\eta}{-V(\eta) + s_l} = t. \quad (24)$$

Let  $t \rightarrow +\infty$ :

$$s_T = \lim_{t \rightarrow \infty} u(x, t)/t = \lim_{t \rightarrow \infty} y(t)/t = 1/E[(-V(\cdot) + s_l)^{-1}], \quad (25)$$

almost surely for a stationary ergodic process  $V$ . Bounding harmonic mean by arithmetic mean:  $s_T \leq s_l - E[V]$ .

Front speed slows down if  $E[V] = 0$ .



$-V + s_l$  changes sign:

$-V + s_l$  has discrete zeros where  $-V'$  changes sign alternately. These zeros are stable and unstable equilibria of characteristic eqn (23).

Then  $y(t)$  will converge to a nearest stable equilibrium from starting point  $x$ . Implying that  $y(t)$  is uniformly bounded in  $t$ .

Front asymptotic speed  $s_T = 0$ , *front trapping* occurs!

In contrast, KPP fronts slow down in gradient flows but are never trapped (Nolen-Xin, 2009).

**Divergence in Shear flow**  $V = (v(y_2), 0)$ ,  $(y_1, y_2) \in \mathbb{R}^2$ ,  $v(y_2)$  is a mean zero stationary ergodic process. Choose control as:

$$\begin{aligned}\dot{y}_1 &= -v(y_2) + s_l - s_l \epsilon (\xi - x_2) e^{-\epsilon s_l t}, \\ \dot{y}_2 &= s_l \epsilon (\xi - x_2) e^{-\epsilon s_l t},\end{aligned}$$

where  $\xi$  is a point so that  $-v(\xi) > 0$ ;  $\epsilon \ll 1$ . Then as  $t \rightarrow \infty$ :

$$\begin{aligned}y_2(t) &= x_2 + (\xi - x_2) (1 - e^{-\epsilon s_l t}) \rightarrow \xi, \\ y_1(t) &= x_1 - \int_0^t v(y_2(\tau)) d\tau + s_l t + O(1), \\ &= -v(\xi)t + s_l t + O(1),\end{aligned}\tag{26}$$

implying that  $y_1/t \rightarrow -v(\xi) + s_l$ .

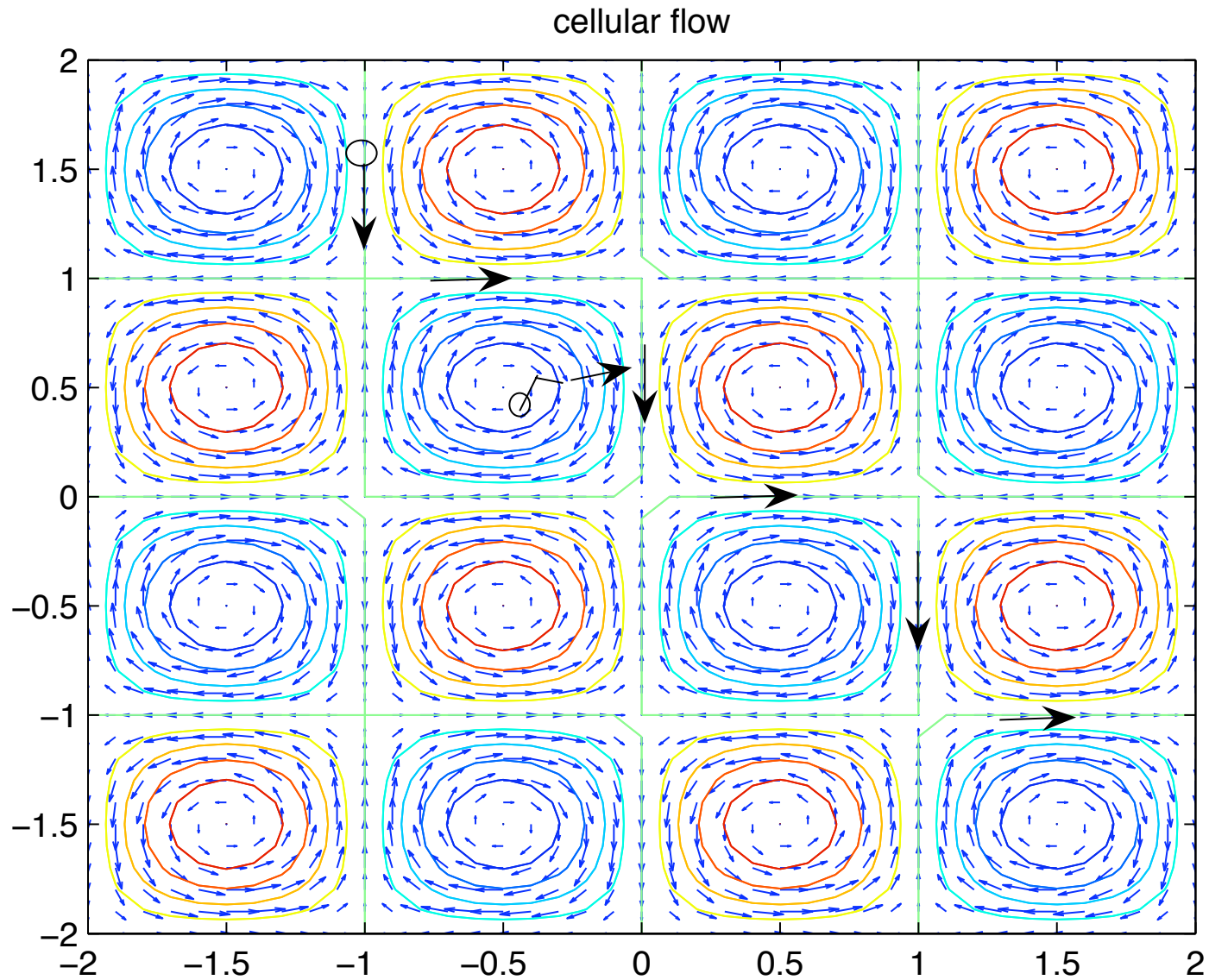
Choose  $\xi$  to reach maximum of  $-v$  on  $[x_2 - L, x_2 + L]$ :

$$s_T \geq s_l + \sup_L \max_{y \in [-L, L]} -v(y).\tag{27}$$

For unbounded Gaussian process  $v$ ,  $s_T$  diverges as  $L \rightarrow \infty$ , same as in quadratic HJ.

## Cell Flows (Hamiltonian Flows with Periodic Array of Vortices)

$$V(x, y) = \text{const } \nabla^\perp H, \quad H = \sin(\pi x) \sin(\pi y).$$



A nearly optimal trajectory to maximize velocity in  $x$  is to follow the flow downward and right along separatrices, and make turns with the help of control. Without the control, the trajectory would get stuck at a saddle point!

The time to go from a small neighborhood of order  $O(q)$  of a saddle point to that of the next saddle point without control is:

$$\int_q^{1-q} dx/A \sin(\pi x) = O(|\log q|/A), \quad A \gg 1, \quad q \ll 1.$$

Choosing  $q = s_L/(A \log A)$ , the control is dominant over cell flow in  $O(q)$  neighborhood of the saddle. So time to make a turn around a saddle is  $O(q/s_L)$ . Total travel time from saddle to saddle is  $O(\log A/A)$ , hence asymptotic velocity is  $s_T = O(A/\log A)$ .

If the starting point  $x$  is not on a separatrix, it takes finite amount of time for the control to move the particle out of the area of closed streamlines to a nearby separatrix:

$$\alpha(s) = -s_l \frac{\nabla \mathcal{H}(y(s))}{|\nabla \mathcal{H}(y(s))|} \cdot \text{sign}(\mathcal{H}(y(s)))$$

until the particle reaches a separatrix then follow the above optimal path to reach the same order of  $s_T$ .

- The enhancement  $O(A/\log A)$  of inviscid G-front speed in cell flows was shown for initial starting point  $x$  on a separatrix in A. Oberman's Thesis (2001) using control representation.
- Abel, Cencini, Vergni, Vulpiani, obtained similar result using formally reduced front equation (2002).
- KPP speeds in cell flow obeys  $A^{1/4}$  (Audoly-Berestycki-Pomeau 2000, Novikov-Ryzhik 2007), Ignition front speeds in cell flow behaves similarly,  $O(A^p)$ ,  $p \in (1/5, 1/4]$  (Kiselev-Ryzhik,2001).
- Lack of dissipation in standard (inviscid) G-equation may be the cause of discrepancy in speed asymptotics ?

## Front Speeds of Viscous G-equation in Cellular Flows:

$$u_t + \delta V(x) \cdot \nabla u = s_l |\nabla u| + \kappa \Delta u, \quad (28)$$

where  $(s_l, \kappa) \ll \delta$ , with Lipschitz continuous initial data.

The traveling front solution of (28) is:

$$u = p \cdot x + H(p) t + w(x), \quad (29)$$

$p$  is a unit vector in  $\mathbb{R}^d$ , and  $w(x)$  is periodic. Equation for  $(H, w(x))$ :

$$H + \delta V(x) \cdot (p + \nabla_x w) = s_l |p + \nabla_x w| + \kappa \Delta_x w, \quad x \in \mathbb{T}, \quad (30)$$

where  $\mathbb{T}$  is the periodic cell  $[0, 2] \times [0, 2]$ .

### Eulerian view:

$$H(p) = s_l \langle |p + \nabla w| \rangle, \quad (31)$$

$L^1$  nonlinearity ! Jensen's inequality gives lower bound:

$$c^* = H(p)/|p| \geq s_l, \quad (32)$$

flow can only enhance the front speed.

**Analysis of  $(H, w)$  equation shows that due to  $L^1$ :**

**Theorem 2 (Nolen-Xin 2009)** *viscous G-front speed  $H(p)$  grows slower than any positive power of  $\delta$  as  $\delta \gg 1$ , if  $s_l \ll \kappa$  independent of  $\delta$ .*

In contrast, viscous quadratic HJ front speed grows at least  $O(\delta^{1/4})$  (Novikov-Ryzhik 2007).

**Lagrangian view:** control representation for the viscous G-equation indicates that large Brownian fluctuation slows down the steering mechanism of control through the vortices. The particle trajectory has a very large probability of falling off a separatrix into rotating vortices and get temporally trapped there. Imagine: in a canoe/kayak slalom race, what a paddler has to do (control) to navigate through a 300m turbulent course to win if he constantly gets a large random kick ?!

**Sketch of Proof of Upper Bound based on Cell Problem:**

$$H^* + AV(x) \cdot (p + \nabla_x w) = R|p + \nabla_x w| + \Delta_x w, \quad x \in \mathbb{T}, \quad (33)$$

where  $A = \delta/\kappa$  and  $R = s_l/\kappa$  and  $H^* = H/\kappa$ . Front speed is  $c^* = H^* \kappa/|p|$ .

As (Novikov-Ryzhik 2007), decompose ( $p = (\lambda, 0)$ ):

$$p \cdot x + w = \lambda x_1 + w = T + S \equiv \zeta, \quad (34)$$

where  $T$  is mean zero and solves the linear inhomogeneous problem:

$$\begin{aligned} \Delta T - AV(x) \cdot \nabla T &= 0, \\ T(x_1 + 2, x_2) &= T(x_1, x_2) + 2\lambda, \quad T(x_1, x_2 + 2) = T(x_1, x_2), \end{aligned} \quad (35)$$

and  $S$ , a mean zero periodic function, solves the nonlinear problem:

$$\Delta S - AV(x) \cdot \nabla S = H^* - R|\nabla(T + S)|. \quad (36)$$

The linear problem (35) is well-studied and  $L^2$  gradient estimate is:

$$C_1 A^{1/2} \lambda^2 \leq \int_{\mathbb{T}} |\nabla T|^2 dx \leq C_2 A^{1/2} \lambda^2, \quad (37)$$

for positive constants  $C_1, C_2$  independent of  $A$ .



**Lemma:** For any power  $\beta > 0$ , there is a constant  $C_\beta$  such that

$$\|\nabla T\|_{L^1(\mathbb{T})} \leq C_\beta(1 + A^\beta)$$

holds for all  $A > 1$ .

Let  $\Omega(h) = \{x \in \mathbb{T} \mid \mathcal{H}(x) < h\}$ ,  $h = \epsilon^p$ ,  $\epsilon = A^{-1}$ , be the boundary region near separatrices. Then  $|\Omega(h)| = O(\epsilon^p \log \epsilon)$ . Cauchy-Schwarz implies for  $q \in (0, p)$ :

$$\int_{\Omega(h)} |\nabla T| dx \leq C_1^{1/2} |\Omega(h)|^{1/2} \epsilon^{-1/4} \leq C \epsilon^{q/2-1/4} \quad (38)$$

Let  $D(h) = \mathbb{T} \setminus \Omega(h)$  denote the interior region, away from the separatrices. For any  $r > 1$ , the energy upper bound:

$$\int_{D(h)} |\nabla T|^2 dx \leq \frac{C_r}{h} \left( \frac{\epsilon}{h^2} \right)^r \quad (39)$$

holds for a positive constant  $C_r$ .

if  $h = \epsilon^p$  for any power  $p < 1/2$ ,

$$\int_{D(\epsilon^p)} |\nabla T|^2 dx \leq C_p \quad (40)$$

for some constant  $C_p$  independent of  $\epsilon \in (0, 1)$ . Thus,

$$\int_{D(\epsilon^p)} |\nabla T| dx \quad (41)$$

is uniformly bounded in  $\epsilon$ . Combining this estimate with (38), we have for any  $p < 1/2$ ,

$$\|\nabla T\|_{L^1(\mathbb{T})} = \int_{D(\epsilon^p)} |\nabla T| dx + \int_{\Omega(\epsilon^p)} |\nabla T| dx \leq C_p + C\epsilon^{q/2-1/4} \quad (42)$$

Choose  $p$  and  $q$  arbitrarily close to  $1/2$ , then for any  $\beta > 0$ ,

$$\|\nabla T\|_{L^1(\mathbb{T})} \leq C_\beta(1 + \epsilon^{-\beta}) = C_\beta(1 + A^\beta) \quad (43)$$

This proves the lemma.

**Proof of Theorem:** function  $\zeta = w + p \cdot x$  increases by  $\lambda$  across the cell in  $x_1$ . It follows:

$$\begin{aligned} H^*(p) &= R \int_{\mathbb{T}} |\nabla \zeta| dx / |\mathbb{T}| \geq R \int_{\mathbb{T}} |\zeta_x| dx / |\mathbb{T}| \\ &\geq R \int_{\mathbb{T}} \zeta_{x_1} dx / |\mathbb{T}| = R \lambda. \end{aligned} \quad (44)$$

By Poincaré inequality and maximum principle, the  $\zeta$  function in the cell  $\mathbb{T}$  satisfies the upper bound:

$$\zeta \leq C[\lambda + \|\nabla \zeta\|_1], \quad (45)$$

implying:

$$\zeta \leq C[\lambda + H^*/R] \leq CH^*/R, \quad (46)$$

for a positive universal constant  $C$ .

The type of inequality (45) also applies to  $T$  and gives:

$$T \leq C[\lambda + \|\nabla T\|_1] \leq C_\beta A^\beta, \quad (47)$$

for all  $A > 1$ . Inequality (47) improves Lemma 1 of (Novikov-Ryzhik 2007) where the upper bound is  $O(\lambda A^{1/4})$ . It follows from (47) and (46) that

$$S \leq C(A^\beta + H^*/R). \quad (48)$$

By (31)-(34):

$$|\mathbb{T}| H^* = R \|\nabla(T + S)\|_1,$$

and so:

$$||\mathbb{T}| H^* - R \|\nabla T\|_1| \leq R \|\nabla S\|_1 \leq 2 R \|\nabla S\|_2. \quad (49)$$

**Gradient estimate of  $S$ :** Multiplying  $S$  to (36), integrating over  $\mathbb{T}$  and applying (48) gives:

$$\begin{aligned}
\|\nabla S\|_2^2 &= R \int_{\mathbb{T}} S |\nabla(T + S)| dx \\
&\leq RC(A^\beta + H^*/R) \int_{\mathbb{T}} |\nabla\zeta| dx \\
&= C(A^\beta + H^*/R)H^*.
\end{aligned} \tag{50}$$

It follows from (49) that:

$$\begin{aligned}
|\mathbb{T}| H^* &\leq 2R \|\nabla S\|_2 + R \|\nabla T\|_1 \\
&\leq 2R \sqrt{C(A^\beta + H^*/R)H^*} + RC(1 + A^\beta).
\end{aligned} \tag{51}$$

Suppose  $H^* > RA^\beta$ . Then (51) implies

$$|\mathbb{T}| H^* \leq R\sqrt{2CH^*H^*/R} + RC(1 + A^\beta) \quad (52)$$

If  $\sqrt{R} = \sqrt{s_l/\kappa}$  is sufficiently small (independently of  $A$ ),

$$H^* \leq RC(1 + A^\beta).$$

Therefore, for any  $\beta > 0$ , there is a constant  $C$  such that

$$H^* \leq RC(1 + A^\beta), \quad (53)$$

holds for all  $A \geq 1$ . Consequently,

$$c^* = H/|p| = H^*\kappa/|p| \leq s_l C \left( 1 + \left( \frac{\delta}{\kappa} \right)^\beta \right) \quad (54)$$

holds for  $\delta > \kappa$ .  $\square$

## Summary of noncoercive HJs and homogenization:

- **In non-coercive HJs due to unbounded random processes:** homogenization may **break down** (shear flows) or persist (gradient flows). A necessary and sufficient condition for homogenization is found for random unbounded Hamiltonian in classical mechanics, namely potential being bounded from above.
- There is **drastically different front speed asymptotics in cell flows of viscous and inviscid G-equations from KPP or ignition reaction or quadratic HJs**. The nonlinearity from  $L^2$  to  $L^1$  makes the key difference.
- **A tutorial presentation of reaction-diffusion, Burgers, HJ fronts in multiscale and random media is in my newly published Springer book at amazon.com for students and researchers at an affordable price \$35.**







## Contents

<b>1</b>	<b>Fronts in Homogeneous Media</b> .....	1
1.1	Traveling Fronts of Burgers and Hamilton–Jacobi Equations .....	5
1.2	Traveling Fronts of Reaction–Diffusion Equations .....	7
1.3	Variational Principles of Front Speeds .....	11
1.4	Random Variables and Stochastic Processes .....	13
1.5	Noisy Burgers Fronts and the Central Limit Theorem .....	18
1.6	Exercises .....	21
<b>2</b>	<b>Fronts in Periodic Media</b> .....	23
2.1	Periodic Media and Homogenization .....	23
2.2	Reaction–Diffusion Traveling Fronts in Periodic Media .....	25
2.3	Existence of Traveling Waves and Front Propagation .....	28
2.4	KPP Fronts and Periodic Homogenization of HJ Equations .....	35
2.5	Fronts in Multiscale Media .....	44
2.6	Variational Principles, Speed Bounds, and Asymptotics .....	48
2.7	Exercises .....	51
<b>3</b>	<b>Fronts in Random Burgers Equations</b> .....	53
3.1	Main Assumptions and Results .....	53
3.2	Hopf–Cole Solutions .....	56
3.3	Asymptotic and Probabilistic Preliminaries .....	58
3.4	Asymptotic Reductions .....	59
3.5	Front Probing and Central Limit Theorem .....	63
3.6	Exercises .....	67
<b>4</b>	<b>Fronts and Stochastic Homogenization of Hamilton–Jacobi Equations</b> .....	69
4.1	Convex Hamilton–Jacobi and Variational Formulas .....	70
4.2	Subadditive Ergodic Theorem and Homogenization .....	73
4.3	Unbounded Hamiltonians: Breakdown of Homogenization .....	78
4.4	Normal and Accelerated Fronts in Random Flows .....	83
4.5	Central Limit Theorems and Front Fluctuations .....	86



x	Contents
4.6	Exercises ..... 90
<b>5</b>	<b>KPP Fronts in Random Media</b> ..... 93
5.1	KPP Fronts in Spatially Random Shear Flows ..... 93
5.2	KPP Fronts in Temporally Random Shear Flows ..... 108
5.3	KPP Fronts in Spatially Random Compressible Flows ..... 117
5.4	KPP Fronts in Space–Time Random Incompressible Flows ..... 123
5.5	Stochastic Homogenization of Viscous HJ Equations ..... 132
5.6	Generalized Fronts, Reactive Systems, and Geometric Models ..... 136
5.7	Exercises ..... 144
	<b>References</b> ..... 145
	<b>Index</b> ..... 157