

*Long-Time Influence of
Small Perturbations.*

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1. Oscillator with One Degree of Freedom

$$\ddot{q}(t) = -F'(q(t)), \quad q(0) = q, \quad \dot{q}(0) = p.$$

$$\begin{cases} \dot{p}(t) = F'(q(t)) \\ \dot{q}(t) = p(t) \end{cases} \quad H(p, q) = \frac{p^2}{2} + F(q)$$

$$x = (p, q) \in \mathbb{R}^2, \quad H(x), \quad \nabla H(x) = \left(-\frac{\partial H}{\partial q}, \frac{\partial H}{\partial p} \right)$$

$$\dot{X}_t = \nabla H(X_t), \quad X_0 = (p, q).$$



$$H(X_t) \equiv H(X_0)$$

$$C(z) = \{x \in \mathbb{R}^2 : H(x) = z\}, \quad G(z)$$

$$T(z) = \oint_{C(z)} \frac{dq}{|\nabla H(x)|} \text{ - period on } C(z)$$

$$m_z(x) = (T(z) |\nabla H(x)|)^{-1}, \quad x \in C(z)$$

- invariant probability measure

$G(z)$ -domain in \mathbb{R}^2 bounded by $C(z)$.

2. Perturbations

$$\dot{X}_t = \nabla H(X_t), X_0 = x; \ddot{Q}(t) = -F'(Q(t))$$

$$\boxed{\dot{\tilde{X}}_t^\varepsilon = \nabla H(\tilde{X}_t^\varepsilon) + \varepsilon \beta(\tilde{X}_t^\varepsilon), \tilde{X}_0^\varepsilon = x.}$$

$$\ddot{Q}_t^\varepsilon = -F_1'(Q(t), Q_{t-\varepsilon}^\varepsilon), Q(0) = q, \dot{Q}^\varepsilon(0) = p$$

$$\ddot{Q}_t^\varepsilon = -F_1'(Q^\varepsilon(t), \Xi_{t/\varepsilon})$$

$$\ddot{Q}^\varepsilon(t) = -F'(Q^\varepsilon(t)), Q^\varepsilon(0) = q + \delta_1, \dot{Q}^\varepsilon(0) = p + \delta_2.$$

stochastic Perturbations

$\Xi(\omega)$ - random variable, $\Xi_t(\omega)$ - stochastic process.

W_t - Wiener stochastic process, \dot{W}_t

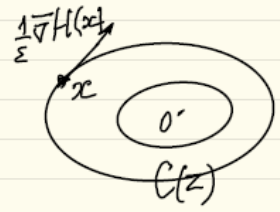
$$\boxed{\dot{\tilde{X}}_t^{\varepsilon, \delta} = \nabla H(\tilde{X}_t^{\varepsilon, \delta}) + \varepsilon \beta(\tilde{X}_t^{\varepsilon, \delta}) + \sqrt{\varepsilon \delta} \sigma(\tilde{X}_t^{\varepsilon, \delta}) * \dot{W}_t}$$

$$\boxed{\tilde{X}_{t/\varepsilon}^{\varepsilon, \delta} = X_t^{\varepsilon, \delta} \quad \dot{X}_t^{\varepsilon, \delta} = \frac{1}{\varepsilon} \nabla H(X_t^{\varepsilon, \delta}) + \beta(X_t^{\varepsilon, \delta}) + \sqrt{\delta} \sigma(X_t^{\varepsilon, \delta}) * \dot{W}_t}$$

$$\mathcal{L} u(x) := \frac{1}{\varepsilon} \nabla H(x) \cdot \nabla u + \beta(x) \nabla u + \frac{\delta}{2} \operatorname{div}(\sigma(x) \nabla u), \sigma(x) = \sigma(x) \sigma^*(x).$$

3. Classical Averaging Principle

$$\dot{X}_t^\varepsilon = \frac{1}{\varepsilon} \nabla H(X_t^\varepsilon) + \beta(X_t^\varepsilon). \quad \text{Fast-Slow components}$$



$$H(X_{t+\Delta}^\varepsilon) - H(X_t^\varepsilon) = \frac{1}{\varepsilon} \int_t^{t+\Delta} \nabla H(X_s^\varepsilon) \cdot \nabla H(X_s^\varepsilon) ds + \int_t^{t+\Delta} \nabla H(X_s^\varepsilon) \cdot \beta(X_s^\varepsilon) ds$$

$$\int_t^{t+\Delta} \nabla H(X_s^\varepsilon) \cdot \beta(X_s^\varepsilon) ds = \Delta \left(\oint_{C(H(X_t^\varepsilon))} \frac{\nabla H(y) \cdot \beta(y)}{T(H(X_t^\varepsilon))} \frac{dl}{|\nabla H(y)|} + \rho \right) + \tilde{\rho}_\Delta$$

$$\frac{1}{T(z)} \oint_{C(z)} \frac{\nabla H(y) \cdot \beta(y)}{|\nabla H(y)|} dl = \frac{1}{T(z)} \int_{G(z)} \operatorname{div} \beta(x) dx = \frac{1}{T(z)} \bar{\beta}(z)$$

$$H(X_t^\varepsilon) \xrightarrow{\varepsilon \downarrow 0} Z_t;$$

$$\dot{Z}_t = \frac{1}{T(Z_t)} \bar{\beta}(Z_t), \quad Z_0 = H(x) \quad /$$

Long-time behavior of X_t^ε for $\varepsilon \ll 1$ can be described by Z_t and a distribution on $C(Z_t)$.

The point 0 (minimum of $H(x)$) is inaccessible in finite time.

4. Stochastic perturbations. One well.

$$\dot{X}_t^{\varepsilon, \delta} = \frac{1}{\varepsilon} \bar{\nabla} H(X_t^{\varepsilon, \delta}) + \beta(X_t^{\varepsilon, \delta}) + \sqrt{\delta} \bar{G}(X_t^{\varepsilon, \delta}) * \dot{W}_t, \quad X_0^{\varepsilon, \delta} = x.$$

$$H(X_t^{\varepsilon, \delta}) \rightarrow Z_t^\delta \text{ as } \varepsilon \downarrow 0.$$

$$\dot{Z}_t^\delta = \frac{\bar{B}(Z_t^\delta)}{T(Z_t^\delta)} + \sqrt{\frac{\delta}{T(Z_t^\delta)}} \bar{G}(Z_t^\delta) * \dot{W}_t, \quad Z_0^\delta = H(x)$$

$$Z_t^\delta \sim L U(z) = \frac{\delta}{2T(z)} \frac{d}{dz} \left(\bar{Q}(z) \frac{dU}{dz} \right) + \frac{1}{T(z)} \bar{B}(z) \frac{dU}{dz}$$

$$\bar{Q}(z) = \int \frac{\operatorname{div}(a(x) \nabla H(x))}{Q(z)} dx, \quad \bar{G}(z) = \sqrt{Q(z)}, \quad \bar{B}(z) = \int \operatorname{div} \beta(x) dx$$

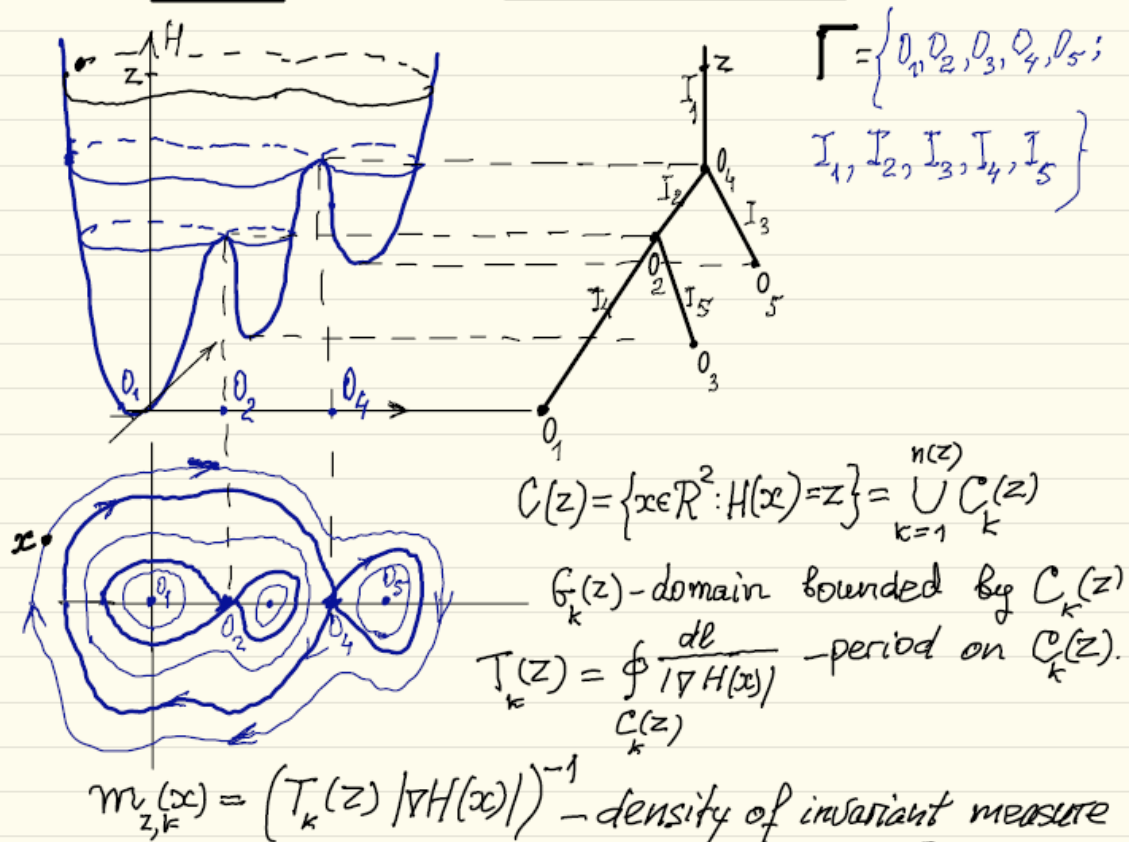
$$T(z) = \int \frac{d\ell}{C(z) |\nabla H(x)|}, \quad G(z)$$

Z_t^δ - diffusion process on $\Gamma = [0, \infty)$.
Point 0 is inaccessible for Z_t^δ .

In the case of one well, the classical averaging principle (extended to stock. perturbations) describes the long-time evolution of the perturbed system.

Remark: If \bar{G} is independent of x , the stochastic integral is defined in the unique way.

5. Multi-well Hamiltonians.

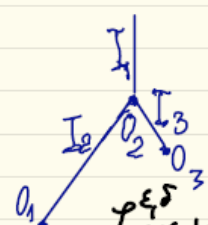


(z, k) -Global coordinate system on Γ :

map $Y: \mathbb{R}^2 \rightarrow \Gamma: Y(x) = (H(x), k(x))$.

$$\boxed{\dot{X}_t = \bar{\nabla} H(X_t), X_0 = x} \quad Y(X_t) \equiv Y(x) = (H(x), k(x)).$$

6. Perturbations of multi-well potential



$$\dot{X}_t^{\varepsilon\delta} = \frac{1}{\varepsilon} \nabla H(X_t^{\varepsilon\delta}) + \beta(X_t^{\varepsilon\delta}) + \sqrt{\delta} \tilde{G} \cdot \dot{W}_t$$

$$\mathcal{L} u(x) = \frac{1}{\varepsilon} \nabla H \cdot \nabla u + \beta \cdot \nabla u + \frac{\delta}{2} \operatorname{div}(\tilde{G} \nabla u),$$

$$a(x) = \theta(x) \theta^*(x).$$



$Y_t^\varepsilon = Y(X_t^{\varepsilon\delta})$ - slow component

Fast component \approx motion along X_t .

Inside an edge I_k : $Y_t^\varepsilon \rightarrow Z_k(t)$

$$\dot{Z}_k^\delta(t) = \frac{1}{T_k(Z_k^\delta(t))} \bar{\beta}_k(Z_k^\delta(t)) + \sqrt{\frac{\delta}{T_k(Z_k^\delta(t))}} \bar{G}_k(Z_k^\delta(t)) * \dot{W}_t$$

$$\bar{\beta}_k(z) = \int \operatorname{div} \beta(x) dx, \quad \bar{G}_k(z) = \sqrt{\bar{a}(z)}, \quad \bar{a}(z) = \int \operatorname{div}(\tilde{G}(x) \nabla H) dx$$

$$Z_k^\delta(t) \sim \mathcal{L}_k^\delta u(z) = \frac{\bar{\beta}_k(z)}{T_k(z)} \frac{du}{dz} + \frac{\delta}{2T_k(z)} \operatorname{div}(\bar{G}_k(z) \nabla u(z)).$$

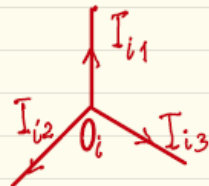
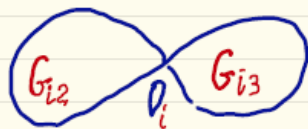
But interior vertices (0_2 in the figure) are accessible in finite time.

Therefore the classical averaging principle should be supplemented by a description what the trajectory is doing after hitting the vertex.

7. Behavior near the vertices.

exterior vertices are inaccessible

Interior vertex 0_i :



When y_t

$$\alpha_{i1} = \alpha_{i2} + \alpha_{i3}; \quad \alpha_{ij} = \int_{G_{ij}} \text{div}(a(x) \nabla H(x)) dx, \quad i=2,3.$$

The limiting process $y_t = (z_t, k_t)$ on Γ is defined in the unique way by the operators L_k and gluing conditions at the interior vertices 0_i :

$$\alpha_{i1} D_1 v(0_i) = \alpha_{i2} D_2 v(0_i) + \alpha_{i3} D_3 v(0_i)$$

where $D_{ij} v(0_i)$ is the first derivative of a function v on Γ along I_{ij} at the vertex 0_i .

Function $v(z, k)$ should be continuous on Γ as well as the function $L_k v(z, k)$.

The operators L_k together with the gluing conditions at interior vertices allow to calculate main term of many interesting characteristics of X_t^ε for $0 < \varepsilon \ll 1$.

8. Modified averaging principle for deterministic perturbations. Stochasticity of systems close to Hamiltonian systems.

$$\dot{X}_t = \nabla H(X_t), \quad X_0 = x \in \mathbb{R}^2$$

$$\dot{\tilde{X}}_t^\varepsilon = \nabla H(\tilde{X}_t^\varepsilon) + \varepsilon \beta(\tilde{X}_t^\varepsilon), \quad X_t^\varepsilon = \tilde{X}_{t/\varepsilon}^\varepsilon,$$

$$\dot{X}_t^\varepsilon = \frac{1}{\varepsilon} \nabla H(X_t^\varepsilon) + \beta(X_t^\varepsilon)$$

Let $\text{div} \beta(x) < 0$, $H(x)$ has saddle points.

$$\dot{X}_t^{\varepsilon, \delta} = \frac{1}{\varepsilon} \nabla H(X_t^{\varepsilon, \delta}) + \beta(X_t^{\varepsilon, \delta}) + \sqrt{\delta} \sigma(X_t^{\varepsilon, \delta}) W_t$$

First, take $\varepsilon \downarrow 0$: $Y(X_t^{\varepsilon, \delta}) \rightarrow Y_t^\delta \sim (L_k^\delta, \text{gluing conditions})$

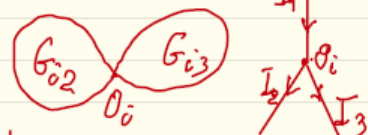
Then take $\delta \downarrow 0$: $Y_t^\delta \rightarrow Y_t = (Z_k(t), k(t))$

deterministic inside the edges (L_k^δ) ;

inside I_k , $\dot{Z}_k = \frac{B_k(Z_k)}{T_k(Z_k)}$.

But when Y_t comes to an interior vertex O_i

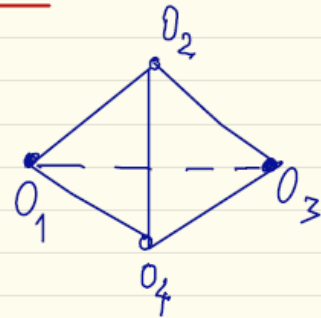
it goes to one of the wells related to O_i with probabilities proportional to



$$\left| \int_{G_{ij}} \text{div} \beta(x) dx \right|, \quad j \in \{2, 3\}.$$

This stochasticity is independent of the type of regularization.

9. Perturbations of systems with asymptotically stable regimes. Large deviations.



$$x_0 = x_0^\varepsilon = x \in \mathbb{R}^n$$

$$\dot{X}_t = b(X_t), \quad \dot{X}_t^\varepsilon = b(X_t^\varepsilon) + \sqrt{\varepsilon} G(X_t^\varepsilon) \dot{W}_t$$

For each fixed T: $\max_{0 \leq t \leq T} |X_t^\varepsilon - X_t| \rightarrow 0$ in probability.

But, if $\sigma(x) = b(x) G^*(x)$ is non-degenerate, X_t^ε sooner or later will deviate of X_t on a distance of order 1.

$$P\left\{ \max_{0 \leq t \leq T} |X_t^\varepsilon - X_t| < \delta \right\} \approx \exp\left\{ -\frac{1}{\varepsilon} S_{OT}(\varphi) \right\}; \quad 0 < \varepsilon, \delta \ll 1.$$

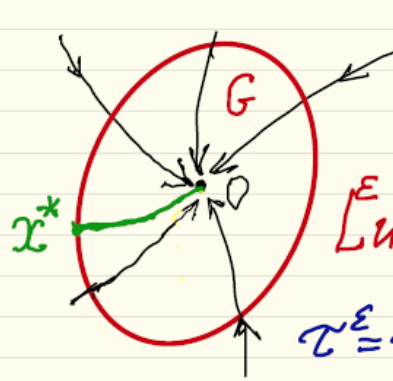
$S_{OT}(\varphi)$ - Action functional.

$$V(x, y) = \inf_{\varphi} \{ S_{OT}(\varphi) : \varphi \in C_{OT}, \varphi_0 = x, \varphi_T = y, T \geq 0 \}$$

Exit problem.

Hierarchy of cycles, Metastability.

10. Exit Problem



$$\dot{X}_t^\varepsilon = b(X_t^\varepsilon) + \sqrt{\varepsilon} \sigma(X_t^\varepsilon) \dot{W}_t, X_0^\varepsilon = x$$

$$a(x) = \sigma(x) \sigma^*(x), x \in \mathbb{R}^n$$

$$L u(x) = \frac{\varepsilon}{2} \sum_{i,j=1}^n b_{ij}(x) \frac{\partial^2 u}{\partial x_i \partial x_j} + \sum_{i=1}^n b_i(x) \frac{\partial u}{\partial x_i}$$

$$\tau^\varepsilon = \min\{t : X_t^\varepsilon \in \partial G\}, X_0^\varepsilon = x \in G.$$

What are $X_{\tau^\varepsilon}^\varepsilon$ -exit position on ∂G ,
Asymptotics of τ^ε and of $E_x \tau^\varepsilon$ as $\varepsilon \rightarrow 0$?

Quasi-potential: $V(x) = V(0, x)$

$$V_0 = \min_{x \in \partial G} V(x) = V(x^*), x^* \in \partial G.$$

$$\lim_{\varepsilon \downarrow 0} \varepsilon \ln \tau^\varepsilon = V_0, \lim_{\varepsilon \downarrow 0} E_x \tau^\varepsilon = V_0$$

If x^* is a unique point of ∂G where $V(x) = V_0$,
then $X_{\tau^\varepsilon}^\varepsilon \rightarrow x^*$ in probability when $\varepsilon \downarrow 0$.

Dirichlet problem: $L u^\varepsilon(x) = 0$ in G , $u^\varepsilon(x) = \psi(x)$.

Then $\lim_{\varepsilon \downarrow 0} u^\varepsilon(x) = \psi(x^*)$

$$u^\varepsilon(x) = E_x \psi(X_{\tau^\varepsilon}^\varepsilon) \rightarrow \psi(x^*) \text{ as } \varepsilon \downarrow 0.$$

11. Hierarchy of cycles. Metastability.

consider a system $\dot{X}_t = b(X_t)$ in \mathbb{R}^n
 with N asymptotically stable regimes.
 Denote them $1, 2, \dots, N$.

Assume that each point $x \in \mathbb{R}^n$, besides
 the points of separatrix surfaces is
 attracted to one of these stable attractors.



17 asymptotically stable regimes:

- 6 cycles of rank 1,
- 3 cycles of rank 2,
- 2 cycles of rank 3,
- 1 cycle of rank 4.

Operator \mathcal{N} (Next):

$\mathcal{N}(i) = j$ such that

$$V_{ij} = \min_{k: k=i, 1 \leq k \leq N} V_{ik}.$$

12. Hierarchy of cycles (generic case).

1-cycle Rotation rate $\tau(C) =$

$$= \max_{i \in C} V_{i, N(i)} = V_{i^*, N(i^*)};$$

$i^* = i^*(C)$ - main state of C;

invariant distribution on C: $m_C^\varepsilon(k)$, $k \in C$,

$$m_C^\varepsilon(k) = A(\varepsilon) \exp \left\{ \frac{1}{\varepsilon} V_{k, N(k)} \right\}, \quad \sum_{k \in C} m_C^\varepsilon(k) = 1.$$

invariant distribution rate:

$$m_C^\varepsilon(k) = \frac{V_{k, N(k)}}{\max_{i \in C} V_{i, N(i)}}$$

Transition probability from $i \in C$ to $j \notin C$

$$\text{rate} \quad \min_{i \in C} [m_C(i) - V_{ij}] = \rho_{C, j}$$

Operator N (next) for 1-cycles: $N(C_1) = C_2$

if C_2 contains k such that

$$\min_{j \notin C_1} \rho_{C_1, j} = \rho_{C_1, k}$$

Exit time (from C) rate $e(C) = \min_{k \in C} \rho_{C, k}$

cycles of rank 2, ...

13. Metastable state (for a given x and λ)

Initial point $x \in \mathbb{R}^n$ is attracted to an asymptotically stable $i(x)$.

$$i(x) = C_{i_0}^{(0)}(x) \in C_{i_1}^{(1)}(x) \in \dots \in C_{i_k}^{(k)}(x) \in C_{i_{k+1}}^{(k+1)}(x) \in \dots \in C = C$$

$$0 < e(C_{i_1}^{(1)}) < \dots < e(C_{i_k}^{(k)}) < e(C_{i_{k+1}}^{(k+1)}) < \dots < \infty$$

If $e_k < \lambda < e_{k+1}$, the process has enough time k to come from x to $C_{i_{k+1}}^{(k+1)}(x)$ and not enough time to leave it.

Then the trajectory spends almost all time near main state $i^*(C_{i_{k+1}}^{(k+1)}(x)) = I(x, \lambda)$

The state (distribution) $I(x, \lambda)$ is called the metastable distribution for given initial point x and time scale $e^{\frac{\lambda}{\varepsilon}} \in (\exp\{\frac{e_k}{\varepsilon}\}, \exp\{\frac{e_{k+1}}{\varepsilon}\})$.

PDE formulation: $\frac{\partial u^\varepsilon(t, x)}{\partial t} = L^\varepsilon u^\varepsilon, u^\varepsilon(0, x) = g(x)$.

$$\lim_{\varepsilon \rightarrow 0} u^\varepsilon(e^{\frac{\lambda}{\varepsilon}}, x) = g(I(x, \lambda))$$

14. General point of view - motion on the simplex of invariant probability measures of the non-perturbed system.

- (i) Finite number of ergodic measures.
- (ii) Exit problem and stopped process
- (iii) Hamiltonian systems with one degree of freedom: graphs give a parametrization of ergodic probability measures.

Landau-Lifshitz equation

$$\dot{X}_t = \nabla F(X_t) \times \nabla G(X_t), \quad X_0 = x \in \mathbb{R}^3, \quad \lim_{|z| \rightarrow \infty} F(z) = \infty.$$

$F(X_t) \equiv F(x)$, $G(X_t) \equiv G(x)$, X_t preserves the volume.

$$S_F(z) = \{x \in \mathbb{R}^3 : F(x) = z\}, \quad S_G(z) = \{x \in \mathbb{R}^3 : G(x) = z\}.$$

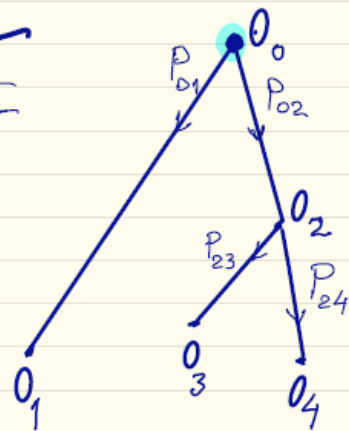
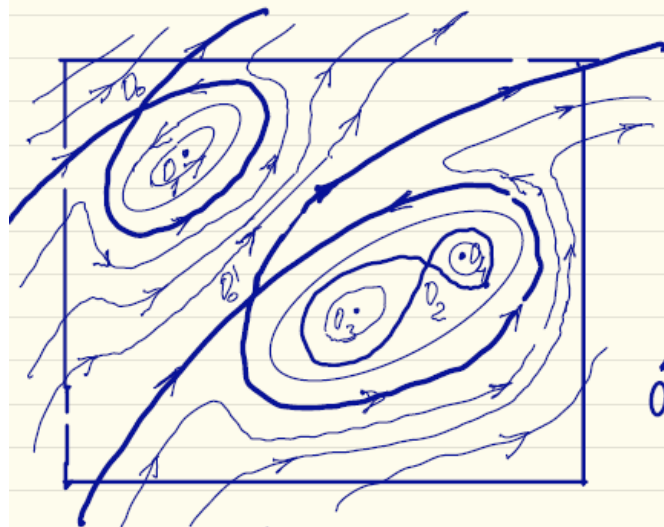
$S_F(z)$ - compact smooth orientable manifold.

Let \mathcal{X} be the genus of $S_F(z)$.

$m_z(x)$ - invariant density on $S_F(z)$: $m_z(x) = \frac{\text{const}}{|\nabla F(x)|}$

If $\mathcal{X} = 0$, the set of invariant measures on $S(z)$ can be described similarly to the F set of invariant measures of a Hamiltonian system with one degree of freedom.

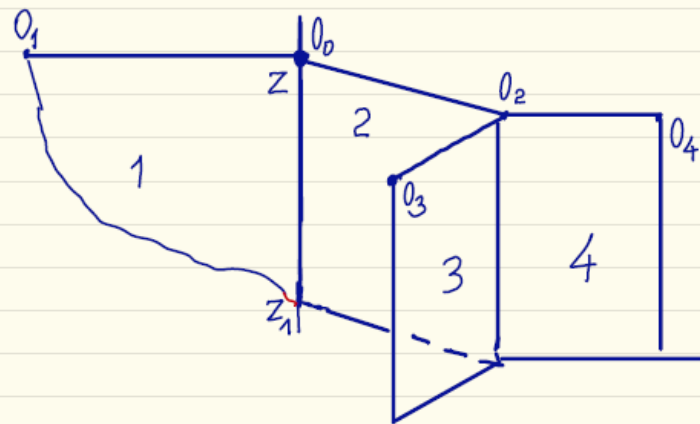
15. Landau-Lifshitz eq. with $\alpha > 0$ ($\alpha=1$).



ergodic probability measures

$S_F(z)$ - Torus ($\alpha=1$).

The whole set of ergodic probability measures can be parametrized by an open book space.



Perturbations

Positive time at O_0 -binding.

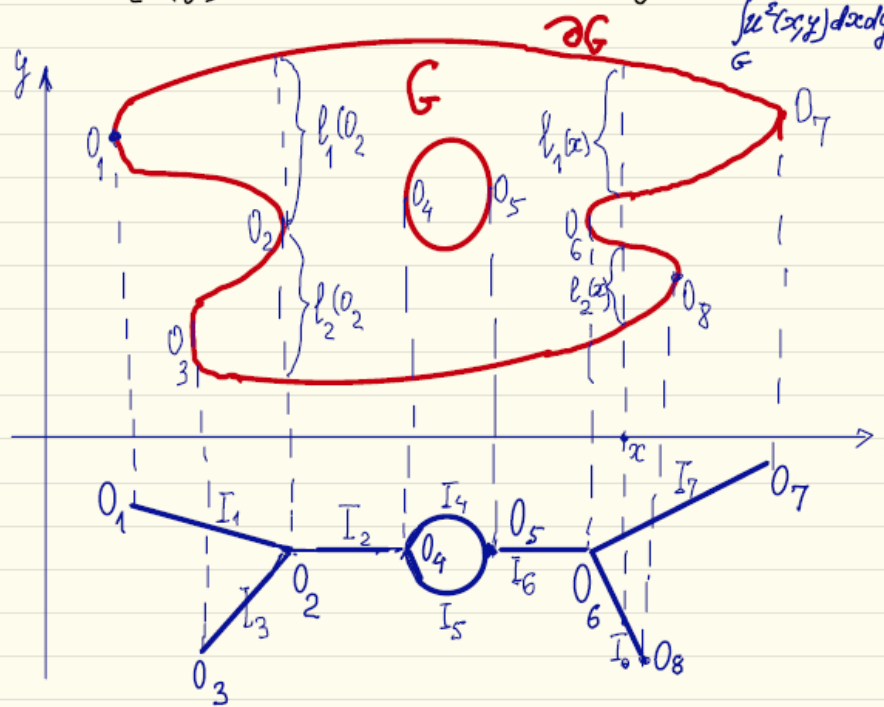
Four pages Open Book.

16. The Neumann problem.

$$\Delta u^\varepsilon(x) = \frac{1}{2} \frac{\partial^2 u^\varepsilon}{\partial x^2} + \frac{1}{2\varepsilon} \frac{\partial^2 u}{\partial y^2} = f(x, y), (x, y) \in G,$$

$$\frac{\partial u^\varepsilon(x, y)}{\partial n_\varepsilon(x, y)} = 0, (x, y) \in \partial G, \int_G f(x, y) dx dy = 0,$$

$$\int_G u^\varepsilon(x, y) dx dy = 0.$$



$$n_\varepsilon(x, y) = (\varepsilon n_1(x, y), n_2(x, y)), (n_1, n_2) \text{ - interior normal}$$

$$\dot{X}^\varepsilon = \sqrt{\varepsilon} \dot{W}_t^1 + \varepsilon n_1(x_t^\varepsilon, y_t^\varepsilon) \dot{\Phi}_t^\varepsilon, \dot{Y}^\varepsilon = \dot{W}_t^2 + n_2(x_t^\varepsilon, y_t^\varepsilon) \dot{\Phi}_t^\varepsilon,$$

$$u^\varepsilon(x, y) = - \int_0^\infty E_{x, y} f(x_{s/\varepsilon}^\varepsilon, y_{s/\varepsilon}^\varepsilon) ds, \quad x, y \in G, \Phi_0^\varepsilon = 0.$$

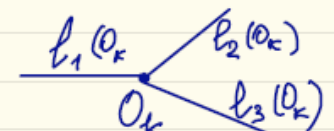
17. Neumann's problem.

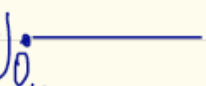
non-perturbed system: $\dot{X}_t = 0, \dot{Y}_t = W_t^2 + m_2(x_t, y_t) \dot{\phi}_t$

limiting process on Γ :

$$X_i(t) \text{ on } I_i \subset \Gamma: X_i(t) \sim \overline{L}_i \sigma(z) = \frac{1}{2l_i(x)} \frac{d}{dx} \left(l_i(x) \frac{d\sigma}{dx} \right)$$

Conditions at the vertices:

(*)  $l_1(O_k) D_1 \sigma(O_k) = l_2(O_k) D_2 \sigma(O_k) + l_3(O_k) D_3 \sigma(O_k)$

(**)  $\lim l_i(x) D_i \sigma(O_k) = 0, \overline{f}_k(x) = \frac{1}{l_k(x)} \int_{l_k(x)} f(x, y) dy$

$$\lim_{\varepsilon \downarrow 0} u^\varepsilon(x, y) = v(x, k(x, y))$$

$$L_k v(x, k) = \overline{f}_k(x), \quad x \in l_k(x)$$

$v(x, k)$ satisfies (*) and (**),

$$\sum_{k: I_k \subset \Gamma} \int_{I_k} v(x, k) l_k(x) dx = 0.$$

∂G .

18. Homogenization

$$\dot{X}_t^\varepsilon = b\left(X_t^\varepsilon, \frac{X_t^\varepsilon}{\varepsilon}\right) + \sigma\left(X_t^\varepsilon, \frac{X_t^\varepsilon}{\varepsilon}\right) \dot{W}_t, \quad X_0^\varepsilon = x \in \mathbb{R}^n$$

$b(x, y), \sigma(x, y)$ are 1-periodic in y .

$$L^\varepsilon = b\left(x, \frac{x}{\varepsilon}\right) \cdot \nabla + \frac{1}{2} \sum_{i,j} a_{ij}\left(x, \frac{x}{\varepsilon}\right) \frac{\partial^2}{\partial x_i \partial x_j}, \quad a(x, y) = \sigma \sigma^*$$

$$\boxed{\tilde{X}_t^\varepsilon = X_{\varepsilon^2 t}^\varepsilon, \quad \tilde{Y}_t^\varepsilon = \varepsilon^{-1} X_{\varepsilon^2 t}^\varepsilon}$$

$$\begin{cases} \dot{\tilde{X}}_t^\varepsilon = \varepsilon^2 b(\tilde{X}_t^\varepsilon, \tilde{Y}_t^\varepsilon) + \varepsilon \sigma(\tilde{X}_t^\varepsilon, \tilde{Y}_t^\varepsilon) \dot{W}_t \\ \dot{\tilde{Y}}_t^\varepsilon = \varepsilon b(\tilde{X}_t^\varepsilon, \tilde{Y}_t^\varepsilon) + \sigma(\tilde{X}_t^\varepsilon, \tilde{Y}_t^\varepsilon) \dot{W}_t \end{cases}$$

Non-perturbed system:

$$\dot{\tilde{X}}_t = 0, \quad \dot{\tilde{Y}}_t = \sigma(\tilde{X}_t, \tilde{Y}_t) \dot{W}_t.$$

For each x , \tilde{Y}_t on T^n has one inv. probab.

measure $m_x(y), y \in T^n$. Put $\bar{b}(x) = \int_{T^n} b(x, y) m_x(y) dy$

$\bar{a}(x) = \int_{T^n} a(x, y) m_x(y) dy$. Then

$$\tilde{X}_{t/\varepsilon^2}^\varepsilon = X_t^\varepsilon \rightarrow \tilde{X}_t \sim \bar{L}u = \bar{b}(x) \cdot \nabla u + \frac{1}{2} \sum_{i,j} \bar{a}_{ij}(x) \frac{\partial^2 u}{\partial x_i \partial x_j}$$

Solutions of initial-boundary and boundary problems for L^ε converge to solutions of corresponding problems for \bar{L} .

19. Reaction-Diffusion Equations

$$\frac{\partial u^\varepsilon(t, x)}{\partial t} = \frac{\varepsilon}{2} \sum_{i,j=1}^n a_{ij}(x) \frac{\partial^2 u^\varepsilon}{\partial x_i \partial x_j} + f(x, u^\varepsilon)$$
$$u^\varepsilon(x) = g(x).$$

non-perturbed problem:

$$\dot{u}(t, x) = f(x, u), \quad u(0, x) = g(x)$$

considered as a flow in the space of functions $g(x) \rightarrow u(t, x)$.

The invariant measures of this flow are δ -functions concentrated on the piece-wise constant functions with values from the set of zeroes of $f(x, u)$.

Under certain conditions one can describe the long-time evolution of these functions under perturbation $\frac{\varepsilon}{2} \sum_{i,j=1}^n a_{ij}(x) \frac{\partial^2}{\partial x_i \partial x_j}$.

20. Perturbations of Markov chains,

Fast oscillating perturbations,
Diffusion approximation,
Small delay, ...