

STOCHASTIC ANALYSIS AND FINSLER MANIFOLDS

Tomasz Zastawniak
University of Hull, UK

I. Diffusion on Finsler manifolds

1. Stochastic development - lifting
Brownian motion to a Finsler manifold
2. Application in mathematical biology:
Random perturbations of Volterra-Hamilton
systems
3. Finslerian diffusion and curvature
4. Laplacian generated by a geodesic
random walk on a Finsler manifold
5. Harmonic forms

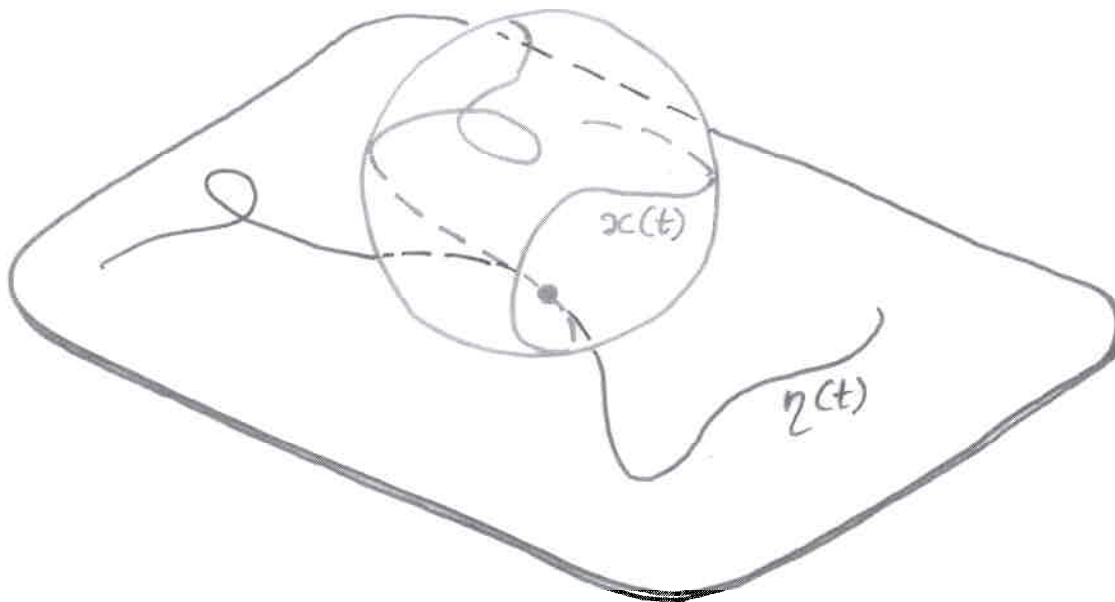
II. Large deviation results involving Finsler metrics

1. Asymptotics of stochastic differential
equations with random coefficients
2. Application in mathematical finance

Stochastic Development

("rolling" a manifold along Brownian motion)

1. Riemannian Manifold M , $\dim M = d$



(a) deterministic case

$\eta(t)$ smooth curve in \mathbb{R}^d

$x(t)$ smooth curve on M

$z(t)$ horizontal lift of $x(t)$ to OM

$$\dot{z} = l(\dot{x})$$

$$l(A)(x, z) = A^i(x) \left(\partial_i - \Gamma_{ij}^k(x) z_n^j \partial_k^n \right)$$

A vect. field on M , $l(A)$ vect. field on OM

$$\begin{cases} \dot{x} = z_i \dot{\eta}^i \\ \dot{z} = l(\dot{x}) \end{cases}$$

development ("rolling")
along η

x independent of z

(b) stochastic case

$w(t)$ Brownian motion in \mathbb{R}^d

$\eta_\pi(t)$ piecewise smooth approximation of $w(t)$

$$\pi : 0 = t_0 < t_1 < \dots < t_n = T$$



$$\begin{cases} \dot{x}_\pi = z_{\pi i} \dot{\eta}_\pi^i \\ \dot{z}_\pi = l(\dot{x}_\pi) \end{cases} \quad \text{development along } \eta_\pi$$

$(x_\pi, z_\pi) \rightarrow (x, z)$ in probability as mesh $\pi \rightarrow 0$

$$\boxed{\begin{cases} dx = z_i \circ dw^i \\ dz = l(\circ dx) \end{cases}}$$

development along BM w

(Stratonovich stochastic differential equations)

Result: $x(t)$ (without $z(t)$!) is Markovian

with generator $D = \frac{1}{2} \Delta$

$$= \frac{1}{2} g^{ij} (z_i z_j - \Gamma_{ij}^k z_k)$$

Def: $x(t)$ called Brownian motion on M

Two extensions of stochastic development to Finsler manifolds

(a) h-stochastic development

$w(t)$ Brownian motion in \mathbb{R}^d

$x(t)$ diffusion on M

$y(t)$ horiz. lift of $x(t)$ to TM

$z(t)$ horiz. lift of $x(t), y(t)$ to OM

$$\begin{cases} dx = z_i \circ dw^i \\ dy = h(\circ dx) \\ dz = l(\circ dy) \end{cases}$$

(b) h \circ -stochastic development

$w(t), v(t)$ independent Brownian motions on \mathbb{R}^d

$x(t), y(t)$ diffusion on TM

$z(t)$ horiz. lift of $x(t), y(t)$ to OM

$$\begin{cases} dx = z_i \circ dw^i \\ dy = h(\circ dx) + z_i \circ dv^i \\ dz = l(\circ dy) \end{cases}$$

Both approaches introduced by
Antonelli, Zastawniak 1993, 1994

Extension to Lagrange manifolds and vector bundles
Antonelli, Hrimiuc 1996

(c) diffusion of vectors, tensors, exterior forms on M

For example, for vectors

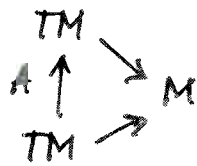
$$\boxed{dy = h(\circ dx)}$$

where

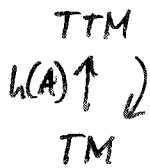
$$h(A)(x, y) = A^i(x) (\partial_i - \Gamma_{ij}^k(x) y^j \dot{\partial}_k)$$

horizontal lift of a vector field A to TM

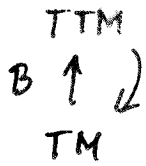
2. Finsler manifold M , $\dim M = d$



$$\begin{aligned} h(A)(x, y) &= A^i(x, y) (\partial_i - F_{ij}^k(x, y) y^j \dot{\partial}_k) \\ &= A^i(x, y) (\partial_i - N_i^k(x, y) \dot{\partial}_k) \\ &= A^i(x, y) \delta_i \end{aligned}$$

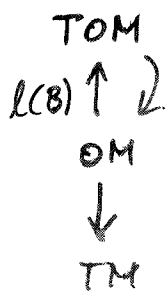


horizontal lift of a Finsler vector field A to a standard vector field $h(A)$ over TM



$B = A^i \delta^i + B^i \dot{\partial}_i$ vector field over TM

$$\begin{aligned} l(B)(x, y, z) &= A^i(x, y) (\delta_i - F_{ij}^k(x, y) z_n^j \partial_k^n) \\ &\quad + B^i(x, y) (\dot{\partial}_i - C_{ij}^k(x, y) z_n^j \partial_k^n) \end{aligned}$$



horizontal lift of a standard vector field B over TM to a vector field $l(B)$ over OM , the Finsler orthonormal frame bundle over TM

Here $(N_i^k, F_{ij}^k, C_{ij}^k)$ Cartan connection
 $x \in M, y \in TM_x, z \in OM_{x,y}$

Application in mathematical biology:

Random perturbations of Volterra-Hamilton systems

N^i size of population of species i

x^i production variable (biomass)

$$\begin{cases} \frac{dx^i}{dt} = N^i & \text{production equation} \\ \frac{dN^i}{dt} = -F_{jk}^i(x, N) \frac{dN^j}{dt} \frac{dN^k}{dt} + \lambda N^i & \text{ecological equation} \end{cases}$$

F_{jk}^i ecological interactions between species

Substitute $N^i = \frac{dx^i}{dt}$ and $s = \frac{1}{\lambda} e^{\lambda t}$ in the second equation to get

$$\frac{d^2 x^i}{ds^2} + F_{jk}^i \frac{dx^j}{ds} \frac{dx^k}{ds} = 0$$

Def: Volterra-Hamilton system of Finsler type

if this is the geodesic equation in a Finsler metric $L(x, y)$

Examples of Volterra-Hamilton systems of Finsler type

$$\begin{cases} \frac{dx^i}{ds} = y^i \\ \frac{dy^i}{ds} + F_{jk}^i(x, y) y^j y^k = 0 \end{cases}$$

Example: Growth of a colony
e.g. two-coral colony

$$L(x, y) = \sqrt{(y^1)^2 + (y^2)^2} e^{\alpha \arctan(y^1/y^2) + (\alpha^2 + 1)(\alpha_1 x^1 + \alpha_2 x^2)}$$

$$\begin{cases} \frac{dy^1}{ds} + 2(\alpha_2 + \alpha\alpha_1) y^1 y^2 + (\alpha_1 - \alpha\alpha_2) [(y^1)^2 - (y^2)^2] = 0 \\ \frac{dy^2}{ds} + 2(\alpha_1 - \alpha\alpha_2) y^1 y^2 + (\alpha_2 + \alpha\alpha_1) [(y^2)^2 - (y^1)^2] = 0 \end{cases}$$

Autonelli Berwald metric

Example: Host-parasite system with social interactions
e.g. rabbit-flea model for myxomatosis

$$L(x, y) = [(y^P)^\mu + (y^H)^\mu]^{1/\mu} e^{\gamma_P x^P - \gamma_H x^H}$$

$$\begin{cases} \frac{dy^P}{ds} = -\gamma_P (y^P)^2 + \gamma_H \frac{\mu}{\mu-1} y^P y^H + \frac{\gamma_P}{\mu-1} \left(\frac{y^H}{y^P}\right)^{\mu-2} (y^H)^2 \\ \frac{dy^H}{ds} = \gamma_H (y^H)^2 + \gamma_P \frac{\mu}{\mu-1} y^P y^H - \frac{\gamma_H}{\mu-1} \left(\frac{y^P}{y^H}\right)^{\mu-2} (y^P)^2 \end{cases}$$

Example: Bacterial colony exchanging chemicals

$$L(x, y) = y^2 \left(\frac{y^2}{y^1}\right)^{1/\lambda} e^{-\alpha_1 x^1 + (\lambda+1)\alpha_2 x^2 + \nu x^1 x^2}$$

$$\alpha_1, \alpha_2, \lambda > 0, \nu \neq 0$$

$$\begin{cases} \frac{dy^1}{ds} + \lambda(\alpha_1 - \nu x^2) (y^1)^2 = 0 \\ \frac{dy^2}{ds} + \lambda\left(\alpha_2 + \frac{\nu}{\lambda+1} x^1\right) (y^2)^2 = 0 \end{cases}$$

Adding noise to a Volterra-Hamilton system

$$\begin{cases} \frac{dx}{ds} = y \\ \frac{dy}{ds} = h\left(\frac{dx}{ds}\right) \end{cases} \quad \begin{array}{l} \text{deterministic} \\ \text{system} \end{array}$$

Rule for noise addition:

The infinitesimal Euclidean distance by which the state (x, y) of the system will be displaced should be proportional to the magnitude of the perturbation dw, dv

- So :
1. Unroll deterministic system to a straight line (constant velocity)
 2. Perturb position and velocity by Brownian motions w and v
 3. Roll back

$$\begin{cases} dx = y ds + z_i^0 dw^i \\ dy = h'(0) dx + z_i^0 dv^i \\ dz = l(0) dy \end{cases} \quad \begin{array}{l} \text{perturbed} \\ \text{system} \end{array}$$

(x, y) Markov diffusion with generator D

$p(s, x, y)$ probability density of $(x(s), y(s))$

$$\begin{aligned} D^* p &= \frac{\partial p}{\partial s} \\ p(0, x, y) &= \varphi(x, y) \\ p(s, x, y) \Big|_{y' y'' = 0} &= 0 \end{aligned}$$

in $y > 0$

initial condition at $s=0$

boundary condition

Solution by Feynman-Kac formula

$$D^* = D + A + V$$

A vector field on TM

V scalar field on TM

$$p(s, x, y) = \mathbb{E}_{s, x, y} \left[\varphi(\tilde{x}(0), \tilde{y}(0)) e^{\int_0^s V(\tilde{x}(u), \tilde{y}(u)) du} \right]$$

where (\tilde{x}, \tilde{y}) diffusion with drift A

$\mathbb{E}_{s, x, y}$ conditional expectation given that
 $\tilde{x}(s) = x, \tilde{y}(s) = y, y' y'' > 0.$

Finslerian Diffusion and Curvature

Berwald space $\delta_l C_{ijk} = 0 \Leftrightarrow D_{jkl}^i = 0 \Leftrightarrow G_{jk}^i$ indep. of y

$C_{ijk}^i = \dot{\partial}_j \dot{\partial}_k (y_{jk}^i y^i y^k)$ Berwald connection coefficients

$D_{jkl}^i = \dot{\partial}_j G_{kl}^i$ Douglas tensor

$C_{ijk} = \frac{1}{2} \dot{\partial}_k g_{ij}$ Cartan torsion tensor

In Berwald spaces there exist normal coordinates (NC)

Def: For h -diffusion $x(t), y(t)$ define

$$U_\varepsilon(x(t), y(t)) = g_{ij}(x(t), y(t)) \mathbb{E}_t \left[x^i(t+\varepsilon) x^j(t+\varepsilon) \right] \text{ in NC}$$

called Quadratic Dispersion

Theorem: In Berwald space

$$U_\varepsilon = \frac{\dim M}{2} \varepsilon + \frac{R}{12} \varepsilon^2 + o(\varepsilon^2)$$

where R is the scalar curvature

(Antonelli, Zastawna 1997)

In Riemannian case $\frac{R}{12}$ is the Onsager-Machlup term
(Onsager, Machlup 1953 ; Takahashi, Watanabe 1981)

Cartan's Lemma for Berwald spaces

$$-g^{jk} \delta_i F_{jk}^i = g^{hk} g^{lm} \delta_h \delta_k g_{ij} = \frac{2}{3} R \quad \text{in NC}$$

Theorem: for h -diffusion $\frac{R}{12}$ replaced by $\frac{R+S}{12}$

Laplacian generated by a geodesic random walk

Riemannian case: Pinsky ~ 1980

M compact Finsler manifold, positive definite metric

$\xi_t(x, y)$ geodesic with $\xi_0(x, y) = x \in M$
 $\dot{\xi}_0(x, y) = y \in IM_x$

$$(T_t^\circ f)(x, y) = f(\xi_t(x, y), \dot{\xi}_t(x, y)), \quad f \in C^\infty(IM)$$

semigroup with generator

$$Z = y^i \partial_i - F_{jk}^i(x, y) y^j y^k \partial_i = y^i \delta_i \quad \text{geodesic flow field on } IM$$

IM_x compact Riemannian manifold with metric induced by the diagonal lift metric
 $g_{ij} dx^i \otimes dx^i + g_{ij} dy^i \otimes dy^i$ on TM

ω_x volume measure on IM_x normalised to 1.

$$(\Pi f)(x) = \int_{IM_x} f(x, y) d\omega_x(y), \quad \Pi: C^\infty(IM) \rightarrow C^\infty(M)$$

projection

e_1, e_2, \dots i.i.d. random variables with exponential distribution $P\{e_n > t\} = \exp(-t)$, $t > 0$

$t_n = e_1 + \dots + e_n$ jump times of a Poisson process

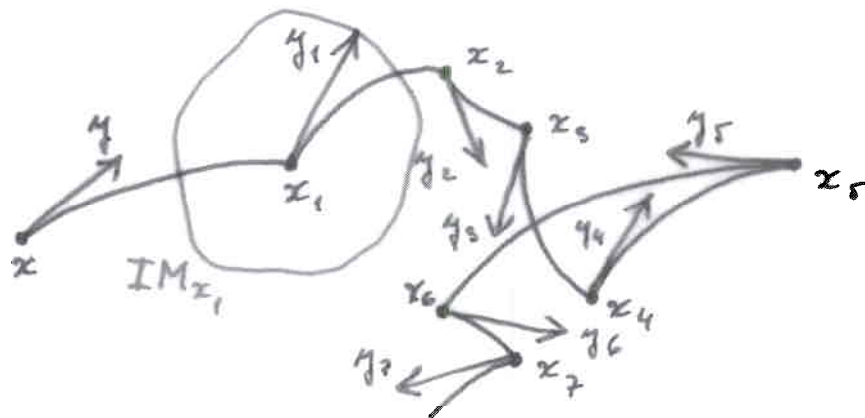
Construction of geodesic random walk

$$x_0 = x, \quad y_0 = y$$

$$x_1 = \xi_{t_1}(x_0, y_0), \quad y_1 \text{ random var. uniformly distr. on } IM_{x_1} \\ \text{independent of } t_1$$

$$x_2 = \xi_{t_2}(x_1, y_1), \quad y_2 \text{ random var. uniformly distr. on } IM_{x_2} \\ \text{independent of } y_1, t_1, t_2$$

⋮



$$X_t(x, y) = \xi_{t-t_n}(x_n, y_n), \quad Y_t(x, y) = \dot{X}_t(x, y) \\ \text{for } t \in [t_n, t_{n+1}), \quad n = 0, 1, 2, \dots$$

$$(T_t f)(x, y) = f(X_t(x, y), Y_t(x, y)), \quad f \in C^\infty(IM)$$

semigroup with generator

$$L f = \lim_{t \rightarrow 0} \frac{T_t f - f}{t} = \mathcal{Z} f + \Pi f - f$$

Central Limit Theorem

$\varepsilon > 0$

$X_t^\varepsilon, Y_t^\varepsilon$ geodesic random walk scaled so that mean time and distance between jumps are ε^2 and ε

(speed $1/\varepsilon$, jump rate $1/\varepsilon^2$)

$(T_t^\varepsilon f)(x, y) = f(X_t^\varepsilon(x, y), Y_t^\varepsilon(x, y))$ semigroup

$L^\varepsilon = \frac{1}{\varepsilon} Z + \frac{1}{\varepsilon^2} (\Pi - 1)$ generator

Theorem: $X_t^\varepsilon \xrightarrow{\text{weakly}} W_t$ as $\varepsilon \searrow 0$

i.e. $\lim_{\varepsilon \searrow 0} E(f(X_t^\varepsilon)) = E(f(W_t)) \quad \forall f \in C^\infty(M)$

where W_t Markov diffusion with generator

$$(\Pi Z^2 f)(x) = \int_{IM_x} g^i g^j (\partial_i \partial_j - F_{ij}^k(x, y) \partial_k) d\omega_x(y)$$

Aronelli, Zastawniak 1998

Def: A -Z Laplacian $\Delta_{AZ} = \Pi Z^2$

$$(\Delta_{AZ} f)(x) = \int_{IM_x} g^i g^j (\partial_i \partial_j - F_{ij}^k(x, y) \partial_k) d\omega_x(y)$$

for $f \in C^\infty(M)$

W_t called Brownian motion on Finsler manifold M

Harmonic Forms

$$H^{ij}(x) = \int_{IM_x} y^i y^j \omega_x \quad \left(= \frac{g^{ij}(x)}{\dim M} \text{ if } M \text{ Riemannian} \right)$$

ω_x normalised volume form on IM_x

scalar product of p -forms φ, ψ

$$\begin{aligned} (\varphi, \psi) &= \int_{IM} \varphi_{i_1 \dots i_p} \psi_{j_1 \dots j_p} H^{i_1 j_1} \dots H^{i_p j_p} \sqrt{g} \omega_x^1 dx^1 \dots dx^n \\ &= \int_M \varphi_{i_1 \dots i_p} \psi_{j_1 \dots j_p} H^{i_1 j_1} \dots H^{i_p j_p} \underbrace{\left(\int_{IM_x} \sqrt{g} \omega_x \right)}_{\sqrt{G}} dx^1 \dots dx^n \end{aligned}$$

$$(d\varphi, \psi) = (\varphi, d^* \psi) \quad \sqrt{G}$$

Harmonic forms $\varphi \in \mathcal{H}^p M$ $(d^*d + dd^*)\varphi = 0$

Hodge decomposition $\Lambda^p M = \mathcal{H}^p M \oplus d\Lambda^{p-1} M \oplus d^*\Lambda^{p+1} M$

Bao-Lackey:

$$\begin{aligned} (\varphi, \psi) &= \int_{SM} \varphi_{i_1 \dots i_p} \psi_{j_1 \dots j_p} g^{i_1 j_1} \dots g^{i_p j_p} \sqrt{g} \omega_x^1 dx^1 \dots dx^n \\ &= \int_M \varphi_{i_1 \dots i_p} \psi_{j_1 \dots j_p} G^{i_1 \dots i_p j_1 \dots j_p} \sqrt{G} dx^1 \dots dx^n \\ G^{i_1 \dots i_p j_1 \dots j_p} &= \frac{1}{\sqrt{G}} \int_{SM_x} g^{i_1 j_1} \dots g^{i_p j_p} \sqrt{g} \omega_x \end{aligned}$$

Leutner:

$$(\varphi, \psi) = \int_M \varphi_{i_1 \dots i_p} \psi_{j_1 \dots j_p} K^{i_1 j_1} \dots K^{i_p j_p} \omega(x) dx^1 \dots dx^n$$

$$K^{ij}(x) = \frac{\int_{L(x,y) \leq 1} y^i y^j dy}{\int_{L(x,y) \leq 1} dy}, \quad \omega(x) = \frac{\int_{|y| \leq 1} dy}{\int_{L(x,y) \leq 1} dy}$$

dy Euclidean volume form in TM_x

Asymptotics of stochastic differential equations with random coefficients

$$dX = b_{\lambda(t)}(X) dt + \sigma_{\lambda(t)}(X) dW_t$$

W_t Brownian motion in \mathbb{R}^m

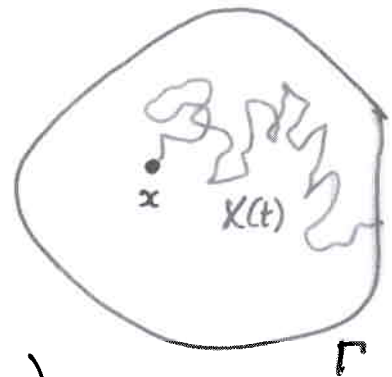
$\lambda(t)$ Markov process with values in state space $\Lambda = \{1, \dots, k\}$ assumed "connected"

$b_\lambda, \sigma_\lambda$ Lipschitz for each $\lambda \in \Lambda$ with values in \mathbb{R}^n and $\mathbb{R}^{n \times m}$

$a_\lambda = \frac{1}{2} \sigma_\lambda \sigma_\lambda^*$ positive definite

$D \subset \mathbb{R}^n$ domain with boundary Γ

initial cond $X(0) = x \in D$
 $\lambda(0) = \lambda \in \Lambda$



$u_\lambda(t, x) = \text{Prob}(X \text{ hits } \Gamma \text{ before time } t)$

Theorem (large deviations result, Lasry + Lions 1995)

$$-2t \log u_\lambda(t, x) \rightarrow L^2(x) \quad \text{as } t \rightarrow \infty$$

uniformly on compact sets in D

where $L(x) = \inf \{ L(x, y) : y \in \Gamma \}$

$$L(x, y) = \inf_{\lambda \in \Lambda} L_\lambda(x, y) \quad \text{Fischer metric}$$

$L_\lambda(x, y)$ Riemannian metric with metric tensor a_λ^{-1}

Theorem (classical large deviations result, Varadhan 1967)

For $dX = b(X)dt + \sigma(X)dW_t$

$$-2t \log u(t, x) \rightarrow L(x) \quad \text{Riemannian distance to boundary, metric tensor } a^{-1}, \quad a = \frac{1}{2} \sigma \sigma^*$$

Application in mathematical finance

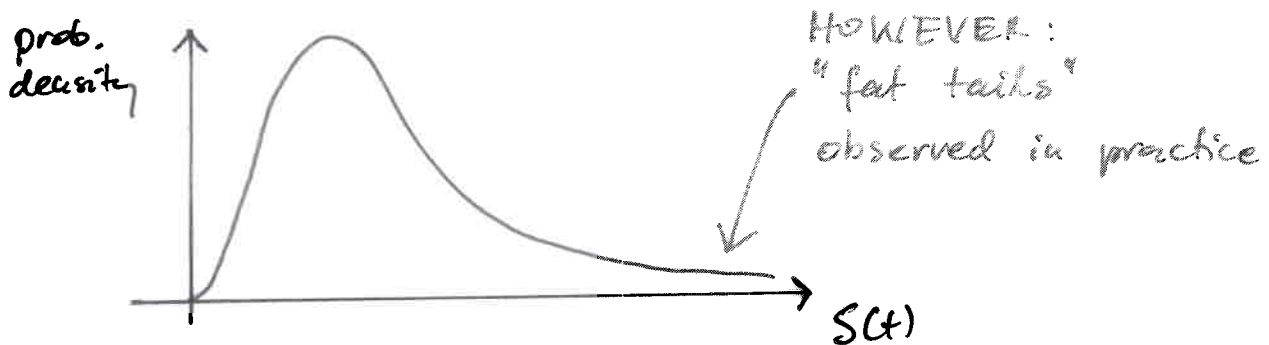
Standard Black-Scholes model of stock price S

$$dS = \mu S dt + \sigma S dW_t$$

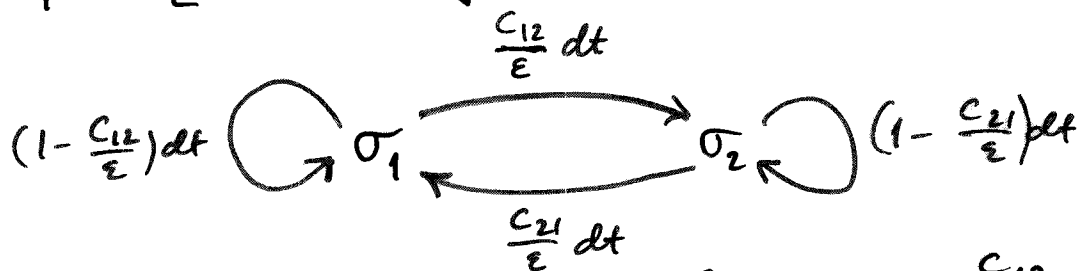
μ rate of return, σ volatility, assumed constant

solution $S(t) = S(0) \exp\left(\mu t - \frac{1}{2}\sigma^2 t + \sigma W_t\right)$

log normal distribution of stock price
(i.e. $\log S(t)$ is normally distributed)



Suppose volatility non-constant, jumping between $\sigma_1 < \sigma_2$ according to Markov process



stationary probability $\frac{c_{21}}{c_{12} + c_{21}}$, $\frac{c_{12}}{c_{12} + c_{21}}$

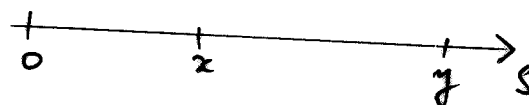
Consider two regimes

1) t fixed
 $\varepsilon \downarrow 0$ frequent jumps regime

"average" volatility observed

$$\bar{\sigma}^2 = \frac{C_{21}\sigma_1^2 + C_{21}\sigma_2^2}{C_{12} + C_{21}}$$

2) ε fixed
 $t \downarrow 0$ large deviations / short time regime



$D = (0, y)$
 $\Gamma = \{y\}$

$$L(x) = L(x, y) = \inf_{\lambda=1,2} L_{\lambda}(x, y)$$

$$L_{\lambda}(x, y) = \inf_{\substack{y(0)=x \\ y(1)=y}} \int_0^1 \sqrt{\frac{1}{\sigma_{\lambda}^2 \dot{y}(t)^2} \dot{y}(t)^2} dt = \frac{1}{\sigma_{\lambda}} |\ln y/x|$$

$$\sigma_1 < \sigma_2 \Rightarrow L_1(x, y) > L_2(x, y)$$

$$u(t, x) \sim e^{-\frac{\ln^2 y/x}{2t\sigma_2^2}}$$

higher volatility σ_2 observed