

Stochastic Calculus, Lecture 12

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Fundamental Solutions, Fokker Planck Equations, Invariant Measure, Conditioning

Let (X_t) be a diffusion process

$$\boxed{dX_t = \sigma(X_t, t) \cdot dW_t + \mu(X_t, t) dt}$$

with associated infinitesimal generator

$$\mathcal{L}\phi = \frac{1}{2}\sigma^2(x, t)\phi_{xx} + \mu(x, t)\phi_x$$

Let $\pi(x, t, y, T) = \text{Prob}\{X_T = y \mid X_t = x\}$

Then

~~$$\frac{\partial \pi}{\partial t} + \mathcal{L}\pi = 0$$~~

$$(1) \quad \begin{cases} \frac{\partial \pi}{\partial t} + \mathcal{L}\pi = 0 & 0 < t < T \\ \pi|_{t=T} = \delta(y-x) \end{cases}$$

This is the backward Fokker Planck equation

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This follows immediately from

$$E[F(X_T) | X_t = x] = \phi(x, t)$$

$$\frac{\partial \phi}{\partial t} + \mathcal{L} \phi = 0$$

If we represent $\phi(x, t)$ as

$$\phi(x, t) = \int F(y) \pi(x, t; y, T) dy$$

$$\int F(y) \left(\frac{\partial}{\partial t} + \mathcal{L}_x \right) \pi(x, t; y, T) dy = 0$$

Since this must hold for all $F(x)$, we conclude that (1) holds.

We now derive a PDE for

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$\pi(x_t; y_T)$ in the "forward variables" (y, T).

We start with the identity

$$\pi(x_t; y, T) = \int \pi(x_t; z, \theta) \pi(z, \theta; y, T) dz$$

(Chapman-Kolmogorov Equation)

which is valid for all θ , $t < \theta < T$

Differentiating wrt θ

$$0 = \int \frac{\partial \pi}{\partial \theta}(x_t; z, \theta) \pi(z, \theta; y, T) dz + \int \pi(x_t; z, \theta) \frac{\partial}{\partial \theta} \pi(z, \theta; y, T) dz$$

$$0 = \int \frac{\partial}{\partial \theta} \pi(x_t, z_t | \theta) \pi(z_t, y_t | \pi) + \int \pi(x_t, z_t) \left(-\frac{1}{2} \sigma^2 \pi_{zz} - \mu \pi_z \right)$$

$$= \int \frac{\partial \pi}{\partial \theta} (x_t, z_t) \pi(z_t, y_t | \pi) +$$

$$+ \int -\frac{1}{2} \sigma^2(z_t) \pi(x_t, z_t) \cdot \pi_{zz}(z_t, y_t | \pi) -$$

(integration by parts) $-\mu(z_t) \pi(x_t, z_t) \pi_z(z_t, y_t | \pi)$

$$= \int \frac{\partial \pi}{\partial \theta} (x_t, z_t) \pi(z_t, y_t | \pi) +$$

$$\int \left[-\frac{1}{2} \left(\sigma^2(z_t) \pi(x_t, z_t) \right)_{zz} + \left(\mu(z_t) \pi(x_t, z_t) \right)_z \right] \pi(z_t, y_t | \pi)$$

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If we set

$$\mathcal{L}_{z,\theta}^* = \left(\frac{1}{2} \sigma^2(z,\theta) \phi(z,\theta) \right)_{zz} - (\mu \phi)_z$$

then

$$0 = \int \left(\frac{\partial \Pi}{\partial \theta} \pi(x_t, z_t, \theta) - \mathcal{L}_{z,\theta}^* \pi(x_t, z_t, \theta) \right) \pi(z_t, \theta, y | \pi) dz$$

Now, let ~~z, \theta~~ $\theta \rightarrow T$

$$0 = \int \frac{\partial \Pi}{\partial T} (x_t, z_t, T) - \mathcal{L}_{z,T}^* \Pi(x_t, z_t, T)$$

$$\frac{\partial \Pi}{\partial T} = \mathcal{L}_{z,T}^* \Pi$$

$\int \delta(z-y) dz$
 (Forward Fokker
Planck Equation)

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The forward FPE describes the evolution of the PDF of the diffusion in the forward variable

Example: $X_t \in O.U.$

$$dX = -kX dt + \sigma dW$$

$$\mathcal{L} \phi = \frac{1}{2} \sigma^2 \phi_{xx} - kX \phi_x$$

$$\mathcal{L}^* \phi = \left(\frac{1}{2} \sigma^2 \phi \right)_{xx} + (kX \phi)_x$$

$$\mathcal{L}^* \phi = \frac{1}{2} \sigma^2 \phi_{xx} + (kX \phi)_x$$

$\pi(x, t; y, T)$ satisfies

$$\frac{\partial \pi}{\partial t} = \mathcal{L}^* \pi$$

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In particular if $T \rightarrow \infty$,

let

$$\bar{\pi}(x, y) = \lim_{T \rightarrow \infty} \pi(x, t; y, T)$$

$$\mathcal{L}_y^* \bar{\pi} = 0$$

$$\frac{1}{2} \sigma^2 \bar{\pi}_{yy} + (Ky \bar{\pi})_y = 0$$

\therefore

$$\frac{1}{2} \sigma^2 \bar{\pi}_{yy} + Ky \bar{\pi} = 0$$

$$\frac{\bar{\pi}_{yy}}{\bar{\pi}} = -\frac{2Ky}{\sigma^2}$$

$$\ln \bar{\pi} = -\frac{Ky^2}{\sigma^2} + C$$

$$\bar{\pi} = \bar{C} e^{-\frac{Ky^2}{\sigma^2}}$$

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$$\bar{\pi}(y) = \frac{1}{\sqrt{2\pi\sigma_e^2}} e^{-\frac{y^2}{2\sigma_e^2}}$$

$$\sigma_e^2 = \frac{\sigma^2}{2K}$$

We recover from the FFPE the long-time behavior of the process X . (Which we knew by explicit solution of SDE!)

Ex 2: CIR Process

$$dX = \alpha \sqrt{X} dW + K(\theta - X) dt$$

$$\mathcal{L}\phi = \frac{1}{2} \alpha^2 X \phi_{XX} + K(\theta - X) \phi_X$$

$$\mathcal{L}^* \phi = \left(\frac{1}{2} \alpha^2 X \phi \right)_{XX} - \left((K(\theta - X)) \phi \right)_X$$

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The equilibrium distribution associated with the CIR process X is $\bar{\pi}(y)$, where

$$\mathcal{L}^X \bar{\pi}(y) = 0$$

$$\left(\frac{1}{2} \alpha^2 y \bar{\pi} \right)_{yy} - \left((k(\theta - y)) \bar{\pi} \right)_y = 0$$

$$\left(\frac{1}{2} \alpha^2 y \bar{\pi} \right)_y - k(\theta - y) \bar{\pi} = 0$$

$$\left(y \bar{\pi} \right)_y = \frac{2k}{\alpha^2} (\theta - y) \bar{\pi}$$

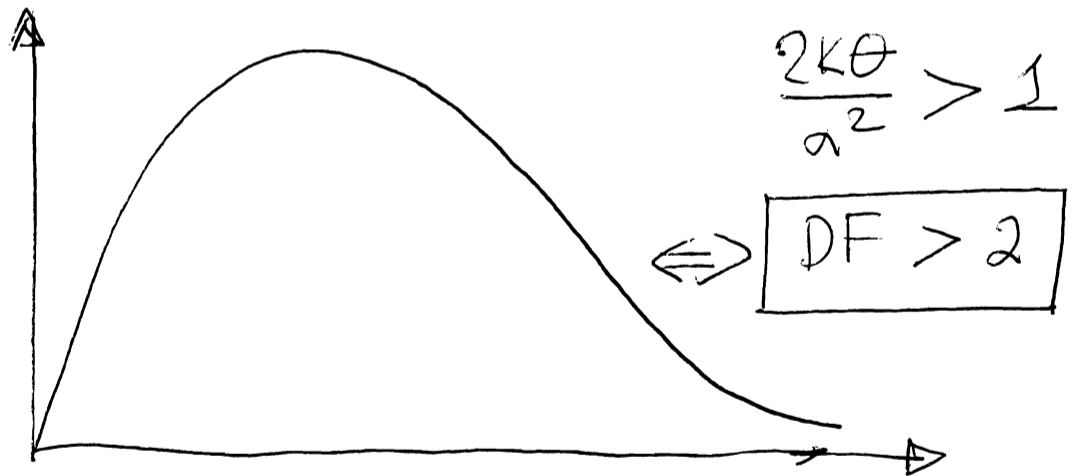
$$= \frac{2k\theta}{\alpha^2 y} \left(\frac{y}{\theta} \bar{\pi} \right) - \frac{2k}{\alpha^2} y \bar{\pi}$$

$$\frac{(y \bar{\pi})_y}{y \bar{\pi}} = \frac{2k\theta}{\alpha^2 y} - \frac{2k}{\alpha^2}$$

$$\ln y \bar{\pi} = \frac{2k\theta}{\alpha^2} \ln y - \frac{2k}{\alpha^2} y + C$$

$$\therefore y \bar{\pi} = C \cdot y^{\frac{2k\theta}{\alpha^2}} e^{-\frac{2k}{\alpha^2} y}$$

$$\bar{\pi}(y) = e y^{\frac{2k\theta}{\alpha^2} - 1} e^{-\frac{2k}{\alpha^2} y}$$



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This is a so-called Gamma distribution.

The constant of integration can be computed easily from

$$\Gamma(p) = \int_0^{\infty} x^p e^{-x} dx$$

$$\Gamma(0) = \int_0^{\infty} e^{-x} dx = 1$$

$$\begin{aligned} \Gamma(1) &= \int_0^{\infty} x e^{-x} dx = -x e^{-x} \Big|_0^{\infty} + \int_0^{\infty} e^{-x} dx \\ &= 1 \end{aligned}$$

$$\Gamma(n) = n!$$

$$\int_0^{\infty} y^{\frac{2k\theta}{\alpha^2} - 1} e^{-\frac{2k}{\alpha^2} y} dy =$$

$$\left(\frac{\alpha^2}{2k} \right) \frac{2k\theta}{\alpha^2} \int_0^{\infty} z^{\frac{2k\theta}{\alpha^2} - 1} e^{-z} dz$$

$$= \left(\frac{\alpha^2}{2K} \right)^{\frac{2K\theta}{\alpha^2}} \Gamma \left(\frac{2K\theta}{\alpha^2} - 1 \right)$$

$$\bar{\pi}(y) = \frac{1}{\left(\frac{\alpha^2}{2K} \right)^{\frac{2K\theta}{\alpha^2}} \Gamma \left(\frac{2K\theta}{\alpha^2} - 1 \right)} y^{\frac{2K\theta}{\alpha^2} - 1} e^{-\frac{\alpha^2}{2K} y}$$

Notice that if $\frac{2K\theta}{\alpha^2} \leq 1$, the boundary conditions which we need to ~~derive~~ derive the FP equation are not satisfied. This means that we need to impose further conditions on the SDE near $x=0$.

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The Brownian Bridge

$B_{t,T}$ = Brownian motion conditional
on $W_T = 0$.

$$\begin{aligned} P_0 [W_t \in A \mid W_T = 0] &= \\ &= \frac{P [W_t \in A ; W_T = 0]}{P [W_T = 0]} \\ &= \frac{\int_A \frac{e^{-\frac{x^2}{2t}}}{\sqrt{2\pi t}} \frac{e^{-\frac{x^2}{2(T-t)}}}{\sqrt{2\pi(T-t)}} dx}{\frac{1}{\sqrt{2\pi T}}} \end{aligned}$$

$$= \int_A e^{-\frac{1}{2}x^2 \left(\frac{1}{t} + \frac{1}{T-t} \right)} \frac{dx}{\sqrt{2\pi \frac{(T-t)t}{T}}} \quad (14)$$

$$= \int_A e^{-\frac{1}{2}x^2 \frac{T}{t(T-t)}} \frac{dx}{\sqrt{2\pi \frac{(T-t)t}{T}}}$$

$$\sigma^2(B_{t,T}) = \frac{t(T-t)}{T} = \frac{t}{T} \left(1 - \frac{t}{T} \right)$$

What about transition?

$$P_{XS} [W_t \in A \mid W_T = 0] =$$

$$\frac{P_{XS} [W_t \in A, W_T = 0]}{P_{XS} [W_T = 0]}$$

$$P_{XS} [W_T = 0]$$

$$= \frac{\int_A e^{-\frac{(y-x)^2}{2(t-s)}} e^{-\frac{y^2}{2(T-t)}} dy}{2\pi\sqrt{(t-s)(T-t)}} \quad (15)$$

$$\frac{e^{-\frac{x^2}{2(T-s)}}}{\sqrt{2\pi(T-s)}}$$

$$\pi_T(x, s, y, t) = \frac{1}{\sqrt{2\pi \frac{(t-s)(T-t)}{(T-s)}}} e^{-\frac{(y-x)^2}{2(t-s)} - \frac{y^2}{2(T-t)} + \frac{x^2}{2(T-s)}}$$

~~$$\pi_T(x, s, y, t) = \frac{\pi(x, s, y, t) \pi(y, t, 0, T)}{\pi(x, s, 0, T)}$$~~

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$$P[B_{t,T} \in A] = \frac{P[W_t \in A; W_T = 0]}{P[W_T = 0]}$$
$$= \lim_{\varepsilon \downarrow 0} \frac{E[W_t \in A; e^{-\frac{1}{2\varepsilon} W_T^2}]}{E[\cancel{W_t} e^{-\frac{1}{2\varepsilon} W_T^2}]}$$

$$V(x,t) = \frac{1}{2\varepsilon} \delta(t-T) x^2$$

$$f(x,t) = E_{x,t} \left[e^{-\frac{1}{2\varepsilon} W_T^2} \right]$$

$$= \int_{-\infty}^{+\infty} e^{-\frac{y^2}{2\varepsilon} - \frac{(y-x)^2}{2(t-t)}} \frac{dy}{\sqrt{\pi(t-t)}}$$

$$\frac{\partial \Sigma_\epsilon(x,t)}{\partial x} \frac{1}{\Sigma_\epsilon(x,t)} = \frac{1}{\sqrt{2\pi\epsilon}} \int \left[\frac{y-x}{2(T-t)} \right] e^{-\frac{y^2}{2\epsilon} - \frac{(x-y)^2}{2(T-t)}} dy$$

$$= \frac{1}{\sqrt{2\pi\epsilon}} \int e^{-\frac{y^2}{2\epsilon} - \frac{(x-y)^2}{2(T-t)}} dy$$

$$\int \frac{y-x}{2(T-t)} f(y) e^{-\frac{(x-y)^2}{2(T-t)}}$$

→
ε ↓ 0

$$= \frac{-\frac{x}{2(T-t)} e^{-\frac{x^2}{2(T-t)}}}{e^{-\frac{x^2}{2(T-t)}} - \frac{-x}{2(T-t)}}$$

This implies that the Brownian Bridge satisfies the diffusion equation

$$dX_t = \frac{-X_t}{(T-t)} dt + dW_t$$

Thus, the BB is an A-R(1) process with a speed of mean-reversion that increases as $t \rightarrow T$.

The FFBB equation is

$$\frac{\partial \pi(x_t, y, t')}{\partial t'} = \frac{1}{2} \pi(x_t, y, t')_{xx} + \left(\frac{y}{(T-t)} \pi(x_t, y, t') \right)_y$$

Conditioning a BM to stay positive for all times

$$P[X_t \in A] = \frac{P[\cancel{W_t} \in A; \min_{s \leq t} W_s > 0]}{P[\min_{s \leq t} W_s > 0]}$$

$$= \frac{P[W_t \in A; \min_{s \leq t} W_s > 0; \min_{t < s \leq T} W_s > 0]}{P[\min_{s \leq T} W_s > 0]}$$

$$= \frac{\int_{\mathbb{R}^+} P[\min_{s \leq t} W_s > 0; W_t = v] P_{y,t}[\min_{t < s \leq T} W_s > 0]}{P_x[\min_{s \leq T} W_s > 0]}$$

$$\left\{ \min_{S \leq T} W_S > 0 \right\} = \left\{ \tau_0 > T \right\}$$

$$\mathbb{1}_{\{\tau_0 > T\}} = \lim_{\varepsilon \downarrow 0} \mathbb{E} \left[\frac{1}{\varepsilon} \int_t^T \chi(W_s) ds \right]$$

Therefore:

$$\zeta(x, t, T) \equiv \mathbb{P}_{x,t}[\tau_0 > T]$$

The conditioned process satisfies

$$dX_t = dW_t + \frac{\zeta_x(x, t, T)}{\zeta(x, t, T)} \cdot dt$$

To calculate $\zeta(x, t, T)$ as $T \rightarrow \infty$ and its logarithmic derivative

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We just ~~say~~ write

$$\frac{\partial \Sigma}{\partial t} + \frac{1}{2} \Sigma_{xx} = 0$$

$$\psi = \ln \Sigma$$

$$\frac{\partial \psi}{\partial t} + \frac{1}{2} (\psi_{xx} + \psi_x^2) = 0$$

$$\rho = \psi_x$$

$$\frac{\partial \rho}{\partial t} + \frac{1}{2} (\rho_{xx} + 2\rho\rho_x) = 0$$

If stationary solution exists

$$\rho_{xx} + 2\rho\rho_x = 0$$

$$\rho_{xx} + (\rho^2)_x = 0$$

~~Ass~~ then

$$\rho_x + \rho^2 = c$$

Since drift should vanish
far away from $x=0$ ^{we have} $c=0$.

$$\rho_x = -\rho^2$$

$$\frac{\rho_x}{\rho^2} = -1 \quad \frac{d}{dx} \left(\frac{1}{\rho} \right) = -X$$

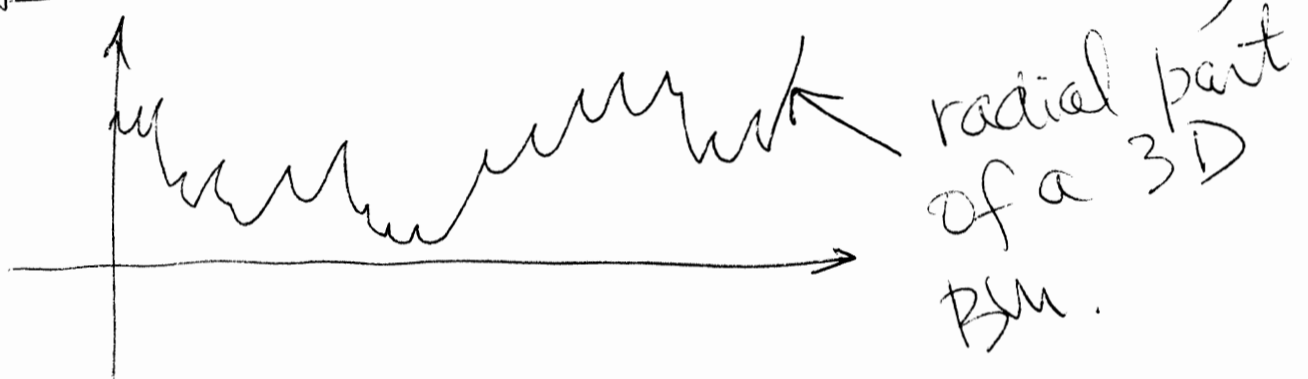
$$\rho = \frac{1}{x}$$

$$dX_t = dW_t + \frac{1}{X_t} dt$$

Recall Itô Bessel process
satisfy.

$$dX_t = dW_t + \frac{n-1}{2X_t} dt$$

Conclusion: Brownian Motion conditioned on never hitting $w=0$ is distributed like a Bessel-3 process (Pitman-Yor, 1970's)



~~BM~~ what about conditioning an Ornstein-Uhlenbeck to remain positive for all times?

0-U.

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$$dX_t = dW_t - k X_t dt$$

New conditional process

$$dX_t = dW_t - k X_t + \frac{\sum x}{\sum} dt$$

$$\frac{\partial S}{\partial t} + \frac{1}{2} \sum_{xx} - k \sum_x = 0$$

"Equilibrium" solution

$$\frac{1}{2} \sum_{xx} = k \sum_x$$

$$\frac{\sum_{xx}}{\sum_x} = 2k$$

$$\ln \sum_x = 2kx$$

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$$\sum_x = \gamma_1 e^{2kx}$$

$$\sum = \gamma_1 \frac{e^{2kx}}{2k} + \gamma_2$$

$$\begin{aligned} \frac{\sum_x}{\sum} &= \frac{\gamma_1 e^{2kx}}{\gamma_1 \frac{e^{2kx}}{2k} + \gamma_2} \\ &= \frac{2k e^{2kx}}{e^{2kx} - \gamma_3} \\ &= \frac{2k}{1 - \gamma_3 e^{-2kx}} \end{aligned}$$

$\frac{\sum_x}{\sum}$

~~It~~ needs to vanish at

$x=0$ to produce infinite drift.

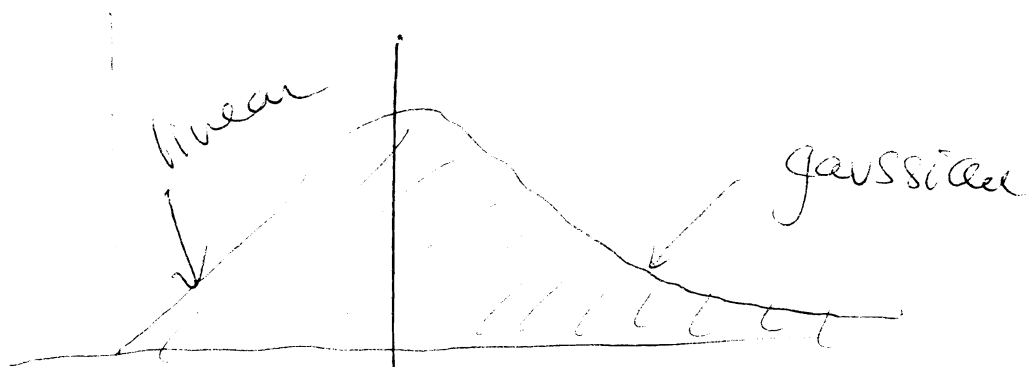
Conclusion: the conditioned

process is

$$dX_t = dW_t - kX_t + \frac{2k}{1 - e^{-2kX_t}} dt$$

It can be shown that this process has invariant measure

$$\bar{\pi}(x) = \frac{1}{Z} e^{-\frac{kx^2}{2}} (e^{-kx})^+$$



The mean is not zero.