

## Existence and Uniqueness Theorem

$\sigma : \mathbb{R}^d \times [0, T] \rightarrow \mathbb{R}^{d \times d}$ ,  $b : \mathbb{R}^d \times [0, T] \rightarrow \mathbb{R}^d$  be Borel measurable,  
 $\exists A < \infty$ ,

$$\|\sigma(x, t)\| + |b(x, t)| \leq A(1 + |x|) \quad x \in \mathbb{R}^d, 0 \leq t \leq T$$

and *Lipschitz*;

$$\|\sigma(x, t) - \sigma(y, t)\| + |b(x, t) - b(y, t)| \leq A|x - y|.$$

$x_0 \in \mathbb{R}^d$  indep of  $B_t$ ,  $E[|x_0|^2] < \infty$ .

Then there exists a unique solution  $X_t$  on  $[0, T]$  to

$$dX_t = b(X_t, t)dt + \sigma(X_t, t)dB_t, \quad X_0 = x_0$$

and  $E[\int_0^T |X_t|^2 dt] < \infty$ .

Uniqueness means that if  $X_t^1$  and  $X_t^2$  are two solutions then

$$P(X_t^1 = X_t^2, 0 \leq t \leq T) = 1$$

## Proof of Uniqueness

Suppose  $X_t^1$  and  $X_t^2$  are solutions

$$X_t^1 - X_t^2 = \int_0^t (b(X_s^1, s) - b(X_s^2, s)) ds + \int_0^t (\sigma(X_s^1, s) - \sigma(X_s^2, s)) dB_s + x_0^1 - x_0^2$$

$$E[|X_t^1 - X_t^2|^2] \leq 4E\left[\left|\int_0^t (b(X_s^1, s) - b(X_s^2, s)) ds\right|^2\right] + 4E\left[\left|\int_0^t (\sigma(X_s^1, s) - \sigma(X_s^2, s)) dB_s\right|^2\right] + 4E[|x_0^1 - x_0^2|^2]$$

$$E\left[\left|\int_0^t (b(X_s^1, s) - b(X_s^2, s)) ds\right|^2\right] \leq A^2 \int_0^t E[|X_s^1 - X_s^2|^2] ds$$

$$E\left[\int_0^t |\sigma(X_s^1, s) - \sigma(X_s^2, s)|^2 ds\right] = E\left[\int_0^t (\sigma(X_s^1, s) - \sigma(X_s^2, s))^2 ds\right] \leq A^2 \int_0^t E[|X_s^1 - X_s^2|^2] ds$$

Call  $\phi(t) = E[|X_t^1 - X_t^2|^2]$

$$\phi(t) \leq 8A^2 \int_0^t \phi(s) ds + 4\phi(0)$$

$$\Phi(t) = \int_0^t \phi(s) ds$$

$$(e^{-8A^2 t} \Phi(t))' = (\Phi'(t) - 8A^2 \Phi(t))e^{-8A^2 t} \leq 4\phi(0)e^{-8A^2 t}$$

$$e^{-8A^2 t} \Phi(t) \leq 4\phi(0)$$

$$\phi(t) \leq 8A^2 \Phi(t) + 4\phi(0) \leq 8e^{8A^2 t} \phi(0)$$

$$E[|X_t^1 - X_t^2|^2] \leq 8e^{8A^2 t} E[|x_0^1 - x_0^2|^2]$$

For each  $0 \leq t \leq T$ ,  $X_t^1 = X_t^2$  a.s. so  $X_t^1 = X_t^2$  for all rational  $t \in [0, T]$  a.s. By continuity this implies that  $X_t^1 = X_t^2$  for all  $t \in [0, T]$  a.s.

## Proof of Existence

$$X_0(t) \equiv x_0$$

$$X_n(t) = x_0 + \int_0^t \sigma(s, X_{n-1}(s)) dB(s) + \int_0^t b(s, X_{n-1}(s)) ds$$

$$E\left[ \sup_{0 \leq t \leq T} |X_n(t) - X_{n-1}(t)|^2 \right]$$

Doob's inequality

$$\leq 4E\left[ \int_0^T \|\sigma(s, X_{n-1}(s)) - \sigma(s, X_{n-2}(s))\|^2 ds \right]$$

$$+ TE\left[ \int_0^T |b(s, X_{n-1}(s)) - b(s, X_{n-2}(s))|^2 ds \right]$$

$$\leq C \int_0^T E[|X_{n-1}(s) - X_{n-2}(s)|^2] ds$$

$$\leq CTE\left[ \sup_{0 \leq t \leq T} |X_{n-1}(t) - X_{n-2}(t)|^2 \right]$$

## Proof.

$$E\left[\sup_{0 \leq t \leq T} |X_n(t) - X_{n-1}(t)|^2\right] \leq CTE\left[\sup_{0 \leq t \leq T} |X_{n-1}(t) - X_{n-2}(t)|^2\right]$$

$$E\left[\sup_{0 \leq t \leq T} |X_n(t) - X_{n-1}(t)|^2\right] \leq \frac{(CT)^n}{n!}$$

$$P\left(\sup_{0 \leq t \leq T} |X_n(t) - X_{n-1}(t)| > \frac{1}{2^n}\right) \leq 2^{2n} E\left[\sup_{0 \leq t \leq T} |X_n(t) - X_{n-1}(t)|^2\right]$$

summable

$$\text{Borel - Cantelli} \Rightarrow P\left(\sup_{0 \leq t \leq T} |X_n(t) - X_{n-1}(t)| > \frac{1}{2^n} \text{ i.o.}\right) = 0.$$

Hence for almost every  $\omega$ ,  $X_n(t) = X_0(t) + \sum_{j=0}^{n-1} (X_{j+1}(t) - X_j(t))$  converges uniformly on  $[0, T]$  to a limit  $X(t)$  which solves the required stochastic integral equation □

Lipschitz condition is *not* necessary

## Theorem

Let  $d = 1$  and

$$\begin{aligned} |b(t, x) - b(t, y)| &\leq C|x - y| \\ |\sigma(t, x) - \sigma(t, y)| &\leq C|x - y|^\alpha, \quad \alpha \geq 1/2 \end{aligned}$$

Then there exists a solution of  $dX_t = b(t, X_t)dt + \sigma(t, X_t)dB_t$  and it is unique

But you do need *some* regularity

$\sigma(x) = \text{sgn}(x)$  and  $dX = \sigma(B)dB$  Not a stochastic differential equation

But  $X$  is a Brownian motion  $dB = \sigma(B)dX$  *is* a stochastic differential equation

But also  $d(-B) = \sigma(-B)dX$  so no uniqueness

# Markov property

$X_t$  can be obtained by solving the stochastic differential equation up to time  $s < t$  and then solving in  $[s, t]$  with initial condition  $X_s$

By uniqueness this gives the same answer

Define the transition probability

$$p(s, x, t, A) = P(X_t^{s,x} \in A)$$

where  $X_t^{s,x}$  is the solution starting at  $x$  at time  $s$

From the construction we have

$$P(X_t^{0,x} \in A \mid \mathcal{F}_s) = p(s, X_s^{0,x}, t, A)$$

which is the Markov property

# Diffusions

A diffusion is a Markov process with transition probabilities  $p(s, x, t, dy)$  satisfying, for each  $\delta > 0$  as  $h \rightarrow 0$ ,

- i.*  $\frac{1}{h} \int_{|y-x| \geq \delta} p(t, x, t+h, dy) \rightarrow 0 \quad \Rightarrow \text{continuous paths}$
- ii.*  $\frac{1}{h} \int_{|y-x| < \delta} (y-x)p(t, x, t+h, dy) \rightarrow b(t, x)$
- iii.*  $\frac{1}{h} \int_{|y-x| < \delta} (y_i - x_i)(y_j - x_j)p(t, x, t+h, dy) \rightarrow a_{ij}(t, x)$



## Formal derivation of the backward equation

$$p(s, x, t, A) = \int p(s, x, s + h, dy) p(s + h, y, t, A)$$

$$0 = \int p(s, x, s + h, dy) \{ p(s + h, y, t, A) - p(s, x, t, A) \}$$

$$0 = \int p(s, x, s + h, dy) \left\{ h \frac{\partial p(s, x, t, A)}{\partial s} + \sum_{i=1}^d (y_i - x_i) \frac{\partial p(s, x, t, A)}{\partial x_i} \right. \\ \left. + \frac{1}{2} \sum_{i,j=1}^d (y_i - x_i)(y_j - x_j) \frac{\partial^2 p(s, x, t, A)}{\partial x_i \partial x_j} + \dots \right\}$$

$$-\frac{\partial p(s, x, t, A)}{\partial s} = \sum_{i=1}^d b_i(t, x) \frac{\partial p(s, x, t, A)}{\partial x_i} + \frac{1}{2} \sum_{i,j=1}^d a_{ij}(t, x) \frac{\partial^2 p(s, x, t, A)}{\partial x_i \partial x_j}$$

## Real derivation

$f(x)$  smooth

$$-\frac{\partial}{\partial s}u = L_s u \quad 0 \leq s < t \quad u(t, x) = f(x)$$

Ito's formula:  $u(s, X(s))$  martingale up to time  $t$

$$u(s, x) = E_{s,x}[u(s, X(s))] = E_{s,x}[u(t, X(t))] = \int f(z)p(s, x, t, dz)$$

Let  $f_n(z)$  smooth functions tending to  $\delta(y - z)$

$$u(s, x) = p(s, x, t, y) \quad \text{if} \quad -\frac{\partial}{\partial s}u = L_s u \quad 0 \leq s < t \quad u(t, x) = \delta(x - y)$$

## Existence result from PDE

Suppose that  $a(t, x)$  and  $b(t, x)$  are bounded and that there are  $\alpha > 0$ ,  $\gamma \in (0, 1]$ ,  $C < \infty$  such that for all  $s, t \geq 0$ ,  $x, y \in \mathbb{R}^d$ ,

- i.  $\xi^T a(t, x) \xi \geq \alpha |\xi|^2$ ,  $\xi \in \mathbb{R}^d$ ,
- ii.  $\|a(s, x) - a(t, y)\| + |b(s, x) - b(t, y)| \leq C(|x - y|^\gamma + |t - s|^\gamma)$ .

Then the backward equation has a solution and furthermore

$$p(s, x, t, A) = \int_A p(s, x, t, y) dy$$

with  $p(s, x, t, y) \geq 0$  jointly continuous in  $s, x, t, y$ . Furthermore,  $p(s, x, t, y)$  is the unique weak solution of the forward equation, i.e.

$$\int f(t, y) p(s, x, t, y) dy - f(s, x) = \int_s^t \int \{\partial_u + \mathcal{L}\} f(u, y) p(s, x, u, y) dy du$$

The solution  $X_t$ ,  $t \geq 0$  of  $dX_t = b(X_t)dt + \sigma(X_t)dB_t$  with  $X_0 = x$  is a Markov process with **infinitesimal generator**

$$L = \frac{1}{2} \sum_{i,j=1}^d a_{ij}(x) \frac{\partial^2}{\partial x_i \partial x_j} + \sum_{i=1}^d b_i(x) \frac{\partial}{\partial x_i}, \quad a = \sigma \sigma^*.$$

Itô's formula

$$\begin{aligned} f(t, X_t) &= f(0, X_0) + \int_0^t \left\{ \partial_s f(s, X_s) + \mathcal{L}f(s, X_s) \right\} ds \\ &\quad + \int_0^t \sum_{i,j=1}^d \sigma_{ij}(s, X_s) \frac{\partial}{\partial x_i} f(s, X_s) dB_s^j \end{aligned}$$

## Example. Brownian motion $d = 1$

$$\mathcal{L} = \frac{1}{2} \frac{\partial^2}{\partial x^2}$$

$$\text{Forward} \quad \frac{\partial p(s, x, t, y)}{\partial t} = \frac{1}{2} \frac{\partial^2 p(s, x, t, y)}{\partial y^2}, \quad t > s$$

$$p(s, x, s, y) = \delta(y - x)$$

$$\text{Backward} \quad - \frac{\partial p(s, x, t, y)}{\partial s} = \frac{1}{2} \frac{\partial^2 p(s, x, t, y)}{\partial x^2}, \quad s < t,$$

$$p(t, x, t, y) = \delta(y - x)$$

## Example. Ornstein-Uhlenbeck Process

$$\mathcal{L} = \frac{\sigma^2}{2} \frac{\partial^2}{\partial x^2} - \alpha x \frac{\partial}{\partial x}$$

$$\text{Forward} \quad \frac{\partial p(s, x, t, y)}{\partial t} = \frac{\sigma^2}{2} \frac{\partial^2 p(s, x, t, y)}{\partial y^2} + \frac{\partial}{\partial y}(\alpha y p(s, x, t, y)), \quad t > s,$$

$$p(s, x, s, y) = \delta(y - x)$$

$$\text{Backward} \quad -\frac{\partial p(s, x, t, y)}{\partial s} = \frac{\sigma^2}{2} \frac{\partial^2 p(s, x, t, y)}{\partial x^2} - \alpha x \frac{\partial p(s, x, t, y)}{\partial x}, \quad s < t,$$

$$p(t, x, t, y) = \delta(y - x)$$

## Under the previous conditions, the following are equivalent

1  $\exists B_t, t \geq 0, dX_t = b(t, X_t)dt + \sigma(t, X_t)dB_t$

2 For each  $\lambda \in \mathbf{R}^d$ ,

$$Z_\lambda(t) = e^{\lambda\{X_t - \int_0^t b(s, X_s)ds\} - \frac{1}{2} \int_0^t \lambda^T a(s, X_s) \lambda ds}$$

is a martingale with respect to  $\mathcal{F}_t$

3 For all smooth  $f(t, x)$ ,

$$f(t, X_t) - \int_0^t \{\partial_s + L\}f(s, X_s)ds$$

is a martingale with respect to  $\mathcal{F}_t$

4 For all smooth  $f(x)$ ,

$$f(X_t) - \int_0^t Lf(X_s)ds$$

is a martingale with respect to  $\mathcal{F}_t$

$$dX_t = \sigma(X_t, t)dW_t + b(X_t, t)dt, \quad X_0 = x$$

$$du(T-t, X_t) = \left\{ -\frac{\partial u}{\partial t} + Lu \right\} (T-t, X_t)dt + \sigma(X_t, t)\nabla u(X_t, t)dB_t.$$

$$\frac{\partial u}{\partial t} = \mathcal{L}u, \quad t > 0 \quad u(0, x) = f(x) \quad \Rightarrow \quad u(T-t, X_t) = \text{martingale}$$

$$E_x[f(X_T)] = u(T, x)$$

## Black-Scholes

Price at time  $t$  of European call option , maturity  $T$  , strike price  $K$

$$V(t, S_t) = e^{-r(T-t)} E_{t, S_t}[(S_T - K)_+]$$

$$dS_t = rS_t dt + \sigma S_t dB_t \quad \text{Geometric Brownian motion}$$

$r$  = riskless interest rate ,  $\sigma$  = stock volatility



$$V(t, S_t) = e^{-r(T-t)} E_{t, S_t}[(S_T - K)_+]$$

$$E_{t, x}[(S_t - K)_+] = \int (y - K)_+ p(T - t, x, y) dy$$

$$\frac{\partial V}{\partial t} = rS \frac{\partial V}{\partial S} - \frac{1}{2} \sigma^2 S^2 \frac{\partial^2 V}{\partial S^2} + rV, \quad V(T, S_T) = (S_T - K)_+$$

$$S_t = S_0 e^{(\mu - \frac{\sigma^2}{2})t + \sigma B_t}$$

$$V(t, S_t) = S_t \Phi \left( \frac{\log \frac{S_t}{K} + (r + \frac{1}{2} \sigma^2)(T - t)}{\sigma \sqrt{T - t}} \right) - e^{-r(T-t)} K \Phi \left( \frac{\log \frac{S_t}{K} + (r - \frac{1}{2} \sigma^2)(T - t)}{\sigma \sqrt{T - t}} \right)$$

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

# Strong Markov Property

Let  $\tau$  be a stopping time

$$f(X_t) - \int_0^t Lf(X_u)du = \text{martingale}$$

$$\begin{aligned} & E[f(X_{t+\tau}) - \int_0^t Lf(X_{u+\tau})du \mid \mathcal{F}_{\tau+s}] \\ &= f(X_{s+\tau}) - \int_0^s Lf(X_{u+\tau})du \quad \text{optional stopping} \end{aligned}$$

$\Rightarrow \tilde{X}_t = X_{\tau+t}, t \geq 0$  is a solution of

$$d\tilde{X}_t = b(t + \tau, \tilde{X}_t)dt + \sigma(t + \tau, \tilde{X}_t)d\tilde{B}_t \quad t \geq 0$$

## Generalized Dirichlet problem.

Let  $D$  be a domain in  $\mathbb{R}^d$ , i.e. a bounded connected open set with a smooth boundary  $\partial D$

Suppose

$$\begin{cases} Lu = 0 & \text{in } D, \\ u = f & \text{on } \partial D \end{cases}$$

$$Lu = \frac{1}{2} \sum_{i,j=1}^d a_{ij}(x) \frac{\partial^2 u}{\partial x_i \partial x_j} + \sum_{i=1}^d b_i(x) \frac{\partial u}{\partial x_i}$$

Let  $X_t$  be the solution of the stochastic differential equation

$$dX_t = \sigma(X_t) dW_t + b(X_t) dt, \quad X_0 = x.$$

Let  $\tau$  be the exit time from the region  $D$

Then  $X_\tau$  is the exit point on the boundary  $\partial D$

$$u(x) = E_x[f(X_\tau)]$$

To show it we cannot use the same conditioning argument as with Brownian motion because we don't have the symmetry anymore

Ito's formula:

$$u(X_{t \wedge \tau}) = \text{martingale}$$

Optional stopping

$$E_x[f(X_\tau)] = E_x[u(X_\tau)] = u(x)$$

# Poisson equation

Suppose

$$\begin{cases} Lu = f & \text{in } D, \\ u = 0 & \text{on } \partial D \end{cases}$$

where  $f$  is some given function defined in  $D$

Ito's formula:

$$u(X_{t \wedge \tau}) - \int_0^{t \wedge \tau} f(X_s) ds = \text{martingale}$$

Optional stopping:  $E_x[u(X_\tau) - u(X_0) - \int_0^\tau f(X_s) ds] = 0$ .

$$u(x) = E_x\left[\int_0^\tau f(X_s) ds\right]$$

Diffusions on manifolds a little tricky to define

### Example. Brownian motion on the circle

If  $(X_1, X_2)$  is a point on the unit circle then the tangent at that point is  $(-X_2, X_1)$ . Therefore it would seem to make sense that the solution of

$$\begin{cases} dX_1 = -X_2 dB \\ dX_2 = X_1 dB \end{cases}$$

would be a Brownian motion on the circle. However we have

$$d(X_1^2(t) + X_2^2(t)) = 2(X_1^2(t) + X_2^2(t))dt$$

so it is not staying on the circle. The true Brownian motion on the circle,

$$(Y_1(t), Y_2(t)) = (\cos(B_t), \sin(B_t))$$

instead satisfies

$$\begin{cases} dY_1 = -\frac{1}{2}Y_1 dt - Y_2 dB \\ dY_2 = -\frac{1}{2}Y_2 dt + Y_1 dB \end{cases}$$

## Martingale representation theorem

$\Omega = C[0, T]$ ,  $\mathcal{F}_T$  = smallest  $\sigma$ -field with respect to which  $B_s$  are all measurable,  $s \leq T$ ,  $P$  the Wiener measure,  $B_t$  = Brownian motion  
 $M_t$  square integrable martingale with respect to  $\mathcal{F}_t$

Then there exists  $\sigma(t, \omega)$  which is

- 1 progressively measurable
- 2 square integrable
- 3  $\mathcal{B}([0, \infty)) \times \mathcal{F}$  mble

such that

$$M_t = M_0 + \int_0^t \sigma(s) dB_s$$