

Backward and Forward Kolmogorov's Eqn

2022年8月3日 18:08

<https://www.math.nyu.edu/~kohn/pde.finance/2015/section1.pdf> pp. 10 - 16

1. Backward, Type 1 PDE / "Expected Payoff"

$$u(t, x) = \mathbb{E}_{x, t} V(x_T) \quad \left\{ \begin{array}{l} u_t + \mathcal{L}u = 0 \\ u(T, x) = V(x) \end{array} \right.$$

2. Forward Kolmogorov's Eqn [Conditional Density] ~ density 固定, 随时间变化. backward 的 $V(x), V(x_1), \dots, V(x_{n+1})$ 向前推多次!

$p(z, s; x, t)$ = probability of being at z at time s , given that it started at x at time t . $\mathbb{P}(X_s = z | X_t = x)$

More precisely: $p(\cdot, s; x, t)$ is the probability density of the state at time s , given that it started at x at time t . Of course p is only defined for $s > t$. To describe a Markov process, p must satisfy the Chapman-Kolmogorov equation

$$p(z, s; x, t) = \int_{\mathbb{R}^n} p(z, s; x_1, t) p(x_1, s; x, t) dx_1$$

for any s_1 satisfying $t < s_1 < s$. Intuitively: the state can get from (x, t) to (z, s) by way of being at various intermediate states z_1 at a chosen intermediate time s_1 . The Chapman-Kolmogorov equation calculates $p(z, s; x, t)$ by adding up (integrating) the probabilities of getting from (x, t) to (z, s) via (z_1, s_1) , for all possible intermediate positions z_1 .

The initial position of a Markov process need not be deterministic. Even if it is (e.g. if $y(0) = x$ is fixed), we may wish to consider a later time as the "initial time." The transition probability determines the evolution of the spatial distribution, no matter what its initial value: if $\rho_0(x)$ is the probability density of the state at time t then

$$\rho(z, s) = \int_{\mathbb{R}^n} p(z, s; x, t) \rho_0(x) dx \quad (12)$$

gives the probability density (as a function of z) at any time $s > t$.

The crucial fact about the transition probability is this: it solves the forward Kolmogorov equation in s and z :

$$-p_s - \sum_i \frac{\partial}{\partial z_i} (f_i(z, s) p) + \frac{1}{2} \sum_{i,j} \frac{\partial^2}{\partial z_i \partial z_j} (g_{ij}(z, s) g_{jk}(z, s) p) = 0 \text{ for } s > t, \quad (13)$$

with initial condition

$$p = \delta_z(z) \text{ at } s = t.$$

We can write the forward Kolmogorov equation as

$$\mathcal{L}^* p + C^* p = 0 \quad \text{w/ } p(t, z, t; x) = \delta_z(x) \quad (14)$$

with

$$\mathcal{L}^* p = - \sum_i \frac{\partial}{\partial z_i} (f_i p) + \sum_{i,j} \frac{\partial^2}{\partial z_i \partial z_j} (a_{ij} p). \quad (15)$$

Here $a_{ij} = \frac{1}{2} (g g^T)_{ij}$ just as before. The initial condition $p = \delta_z(z)$ encapsulates the fact, already noted, that the graph of $p(\cdot, s; x, t)$ becomes infinitely tall and thin at x as s decreases to t . The technical meaning is that

Note: 内积 $\langle \mathcal{L}x, y \rangle = \langle x, \mathcal{L}^* y \rangle$.
 证明: 设 $\mathcal{L}^* = \mathcal{L}^T$

In function space $\langle f, g \rangle = \int_{-\infty}^{\infty} f g dx$

Proof.

$$\begin{aligned} \Rightarrow u(t, x) &= \mathbb{E}_{x, t} V(x_T) = \text{backward} \\ &= \int_{-\infty}^{\infty} V(z) p(T, z, t; x) dz \end{aligned}$$

$$u_t + \mathcal{L}u = 0 \quad \mathbb{E}_{x, t} (V_{X_T}) = \mathbb{E}_{x, t} V(X_T)$$

$$\int_{-\infty}^{\infty} u(s, z) p(s, z, t; x) dz = u(t, x)$$

$$\Rightarrow \langle \mathcal{L}f, g \rangle = \langle f, \mathcal{L}^* g \rangle$$

$$\text{证明 } \mathcal{L}f = a f_x + \frac{1}{2} \sigma^2 f_{xx}$$

$$\mathcal{L}^* g = (a g)_x + \frac{1}{2} (\sigma^2 g)_{xx}$$

$$\begin{aligned} \Rightarrow \int \mathcal{L}f g dx &= \int a f g_x dx + \frac{1}{2} \int \sigma^2 f g_{xx} dx \\ &= f(a g) \Big|_{-\infty}^{\infty} - \int f g_x dx + \frac{1}{2} \left(f(\sigma^2 g) \Big|_{-\infty}^{\infty} - \int f(\sigma^2 g)_{xx} dx \right) \\ &= - \int f g_x dx - \frac{1}{2} \left(\int f(\sigma^2 g)_{xx} dx - \int f(\sigma^2 g)_{xx} dx \right) \\ &= - \int f g_x dx + \frac{1}{2} \int f(\sigma^2 g)_{xx} dx = \int f \mathcal{L}^* g dx \quad \square \end{aligned}$$

$$\Rightarrow u(T, x) - u(t, x) \stackrel{\text{It\^o}}{=} \int_t^T u_s ds + \int_t^T u_s dX_s + \frac{1}{2} \int_t^T u_{ss} (dX_s)^2$$

类似马氏链 finite MC ~ strong probability.

$$\text{例: BM: } p(s, z; t, x) = \frac{1}{\sqrt{2\pi(s-t)}} e^{-\frac{(z-x)^2}{2(s-t)}} \quad \text{若 } x_t, x_s \sim N(0, s-t)$$

$$p(x, s; x, t) = \frac{1}{\sqrt{2\pi(s-t)}} e^{-\frac{(x-x)^2}{2(s-t)}} = \frac{1}{\sqrt{2\pi(s-t)}} \int_{-\infty}^{\infty} \delta(x-z) dz = 1$$

$$\begin{aligned} \int_0^T dt \quad dX_t &= a dt + \sigma^1 dB_t^1 + \dots + \sigma^m dB_t^m \\ dX_t^i &= a^i dt + \sigma^{i1} dB_t^1 + \dots + \sigma^{im} dB_t^m \end{aligned}$$

$$\vdots \\ dX_t^m = a^m dt + \dots + \sigma^{m1} dB_t^1 + \dots + \sigma^{mm} dB_t^m$$

$$\text{Ito's lemma: } dX_t = a dt + \sigma dB_t \Rightarrow p_t - (a(s, z) p)_t + \frac{1}{2} (\sigma^2(s, z) p)_{ss} = 0$$

这是初始 position. 向未来推. Forward.

$$\begin{aligned} \int_0^T u(t, x) &= \mathbb{E}_{x, t} V(X_T) \\ &= \int_{-\infty}^{\infty} V(z) p(T, z, t; x) dz \end{aligned}$$

