

**Problem Set 3**Due at lecture on Thursday, March 31, 2005.

1. **Modified Kramers-Moyall Expansion.** Let  $P_N(x)$  be the probability density for a random walker (or, equivalently, the concentration of a large number of independent walkers) to be at position  $x$  at time  $t_N = N\tau$ . The walker's displacements  $(x, t) \rightarrow (x', t + \tau)$  are independently chosen with a transition probability  $p(x', t + \tau | x, t)$  at regular intervals of time  $\tau$ . Suppose that the moments, which depend on time and space,

$$M_n(x, t, \tau) = \int p(x + y, t + \tau | x, t) y^n dy \quad (1)$$

are finite. Consider a continuous-time probability density (or concentration),  $\rho(x, t)$ , satisfying  $\rho(x, N\tau) = P_N(x)$ .

In class, we formally derived a PDE expansion for  $\rho(x, t)$ :

$$\frac{\partial \rho}{\partial t} + \sum_{n=2}^{\infty} \frac{\tau^{n-1}}{n!} \frac{\partial^n \rho}{\partial t^n} = \sum_{n=1}^{\infty} \left( -\frac{\partial}{\partial x} \right)^n [D_n(x, t, \tau) \rho(x, t)] \quad (2)$$

where  $D_n(x, t, \tau) = M_n(x, t, \tau)/(n! \tau)$ . Without the second term on the left-hand side, this is called the *Kramers-Moyall expansion*.

In the limit  $\tau \rightarrow 0$ , assume that the transition moments are finite and scale like,  $M_1 \sim D_1 \tau$ ,  $M_2 \sim 2D_2 \tau$ , and  $M_n \sim M_2^{n/2} = O(\tau^{n/2})$  for  $n > 2$  (which follows if the CLT holds on very small time scales). At leading order, we have the *Fokker-Planck equation*,

$$\frac{\partial \rho}{\partial t} + \frac{\partial}{\partial x} (D_1 \rho) = \frac{\partial^2}{\partial x^2} (D_2 \rho) + O(\tau^{1/2}) \quad (3)$$

but please calculate all terms up to  $O(\tau)$  in a “modified” expansion of the form

$$\frac{\partial \rho}{\partial t} = \sum_{n=0}^{\infty} \tau^{n/2} L_n \rho \quad (4)$$

where the operators  $L_n$  involve only *spatial* derivatives.

2. **Black-Scholes Formulae for Options Prices.**

- (a) Solve the Black-Scholes equation,

$$\frac{\partial w}{\partial t} + rx \frac{\partial w}{\partial x} + \frac{\sigma^2 x^2}{2} \frac{\partial^2 w}{\partial x^2} = rw \quad (5)$$

backward from maturity,  $t < T$ , for the long position of a call option, with payoff,  $w(x, T) = y(x) = \max(x - K, 0)$ .

- (b) Show that the solution is equivalent to a “risk neutral valuation”,

$$w(x, t) = e^{-r(T-t)} \langle y(x) \rangle \quad (6)$$

where the expectation is taken with the final value of the underlying asset,  $x_T = \int_t^T dx_t$ , given by a geometric Brownian motion, which solves the SDE,

$$dx = rxd t + \sigma x dz \quad (7)$$

with volatility  $\sigma$  and mean return,  $r$ , the risk free rate (not the actual expected return). [Note:  $x_T$  has a lognormal distribution.]

- (c) Let  $w_c(x, t)$  and  $w_p(x, t)$  be the prices of (long) call and put options, respectively, on the same underlying asset, with the same maturity,  $T$ , volatility  $\sigma$ , risk-free rate  $r$ , and strike price,  $K$ . Explain why  $w_p(x, t)$  can be found from your solution above, using “put-call parity”:

$$w_p(x, t) = w_c(x, t) - x + Ke^{-r(T-t)} \quad (8)$$

3. **Continuum Limit of Bouchaud-Sornette Options Theory.** Consider a discrete random walk for an underlying asset with (additive) independent steps. Assume the displacements  $y = \delta x$  in each time step  $\tau$  have low order moments which depend on the current price,

$$\langle \delta x \rangle = \mu x \delta t \quad (9)$$

$$\langle \delta x^2 \rangle = \sigma^2 x^2 \delta t + \mu^2 x^2 \delta t^2 \quad (10)$$

$$\langle \delta x^3 \rangle = \sigma^3 \lambda_3 x^3 \delta t^{3/2} + 3\mu\sigma^2 x^3 \delta t^2 + O(\delta t^3) \quad (11)$$

$$\langle \delta x^4 \rangle = \sigma^4 (\lambda_4 + 3) x^4 \delta t^2 + O(\delta t^{5/2}). \quad (12)$$

As discussed in class this is a general model for random returns in each time step  $\delta t$ .

Assume the Bouchaud-Sornette strategy of minimizing the “quadratic risk” or variance of the return of a position consisting of the option and short  $\phi$  of the underlying. Since minimizing the total variance is equivalent to minimizing the variance in each time step, we get the least-squares fit equations given in class, as a recursion for  $w(x, t)$ .

$$u(x, t) = w(x, t) - \phi(x, t)x = e^{-r\delta t} \left[ \int w(x + \delta x, t + \delta t) p(\delta x, \delta t) d\delta x - \phi(x, t) \int (x + \delta x) p(\delta x, \delta t) d\delta x \right] \quad (13)$$

and

$$\phi(x, t) = \frac{1}{\sigma^2 x^2 \delta t} \int (\delta x - \mu x \delta t) w(x + \delta x, t + \delta t) p(\delta x, \delta t) d\delta x \quad (14)$$

Consider the limit  $\delta t \rightarrow 0$  in these equations and (formally) derive a PDE for  $w(x, t)$  and an expression for  $\phi(x, t)$  accurate to  $O(\delta t^{1/2})$ . Each should involve only  $x$  derivatives of  $w$  (aside from  $\partial w / \partial t$  in the PDE). Recover the Black-Scholes equation at leading order  $O(1)$ . [Extra credit: derive the expansions to  $O(\delta t)$ . Extra extra credit: Try solving them for a call option!]