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END SEM → PAPER-II

(4) PDE modelling Asian option with Arithmetic Avg.
is :

$$\left[\frac{\partial V}{\partial t} + \frac{1}{2} \sigma^2 S^2 \frac{\partial^2 V}{\partial S^2} + rS \frac{\partial V}{\partial S} + S \frac{\partial V}{\partial A} - rV = 0 \right]$$

The Payoff for Arithmetic Avg strike call Asian option is given by .

$$\begin{aligned} \text{Payoff} &= \left(S_T - \frac{1}{T} A_T \right)^+ \quad \text{where } A_t = \int_0^t S_0 d\theta \\ &= S_T \left(1 - \frac{1}{T S_T} \int_0^T S_0 d\theta \right) \end{aligned}$$

Now Define R_t as follows

$$\left[R_t = \frac{1}{S_t} \int_0^t S_0 d\theta \right] \Rightarrow \left[R = \frac{A}{S} \right] \text{ (as given)}$$

Then $\left[\text{Payoff} = S_T \left(1 - \frac{R_T}{T} \right)^+ \right]$

Now Using the transformation that is given

$$V(S, A, t) = S \cdot H(R, t) \quad \text{with } R = A/S$$

we have following

$$\left[\frac{\partial R}{\partial A} = \frac{1}{S} \right] \quad \text{and} \quad \left[\frac{\partial R}{\partial S} = -\frac{A}{S^2} \right]$$

Now

$$\frac{\partial V}{\partial t} = \frac{\partial}{\partial t} (S \cdot H(R, t)) = S \frac{\partial H}{\partial t}$$

$$\frac{\partial V}{\partial S} = \frac{\partial}{\partial S} (S \cdot H(R, t)) = H + S \cdot \frac{\partial H}{\partial S}$$

$$= H + S \cdot \frac{\partial H}{\partial R} \cdot \frac{\partial R}{\partial S}$$

$$= H + S \cdot \left(-\frac{A}{S^2}\right) \cdot \frac{\partial H}{\partial R} \quad \left\{ \text{Using } \frac{\partial R}{\partial S} = -\frac{A}{S^2} \right\}$$

$$\boxed{\frac{\partial V}{\partial S} = H - R \frac{\partial H}{\partial R}} \quad \left\{ \text{Using } R = A/S \right\}$$

$$\frac{\partial^2 V}{\partial S^2} = \frac{\partial}{\partial S} \left(H - R \frac{\partial H}{\partial R} \right) = \frac{\partial}{\partial R} \left(H - R \frac{\partial H}{\partial R} \right) \cdot \frac{\partial R}{\partial S}$$

$$= -\frac{A}{S^2} \left\{ \frac{\partial H}{\partial R} - \frac{\partial}{\partial R} \left(R \cdot \frac{\partial H}{\partial R} \right) \right\} \quad \left(\text{Using } \frac{\partial R}{\partial S} = -\frac{A}{S^2} \right)$$

$$= -\frac{A}{S^2} \left\{ \frac{\partial H}{\partial R} - \frac{\partial H}{\partial R} - R \frac{\partial^2 H}{\partial R^2} \right\}$$

$$= \frac{AR}{S^2} \frac{\partial^2 H}{\partial R^2} \Rightarrow \boxed{\frac{\partial^2 V}{\partial S^2} = \frac{R^2}{S} \frac{\partial^2 H}{\partial R^2}} \quad \left(\text{Using } A = RS \right)$$

$$\frac{\partial V}{\partial A} = \frac{\partial}{\partial A} (S \cdot H) = S \cdot \frac{\partial H}{\partial A} = S \cdot \frac{\partial H}{\partial R} \cdot \frac{\partial R}{\partial A}$$

$$\Rightarrow \boxed{\frac{\partial V}{\partial A} = \frac{\partial H}{\partial R}} \quad \left(\text{Using } \frac{\partial R}{\partial A} = \frac{1}{S} \right)$$

Now substituting all the values in PDE, we get

$$S \frac{\partial H}{\partial t} + \frac{1}{2} \sigma^2 S^2 \frac{R^2}{S} \frac{\partial^2 H}{\partial R^2} + S \left(r \left(H - R \frac{\partial H}{\partial R} \right) + \frac{\partial H}{\partial R} \right) - rSH = 0$$

$$\Rightarrow \frac{\partial H}{\partial t} + \frac{1}{2} \sigma^2 R^2 \frac{\partial^2 H}{\partial R^2} + rH - rR \frac{\partial H}{\partial R} + \frac{\partial H}{\partial R} - rH = 0$$

$$\Rightarrow \left[\frac{\partial H}{\partial t} + \frac{1}{2} \sigma R^2 \frac{\partial^2 H}{\partial R^2} + (1-rR) \frac{\partial H}{\partial R} = 0 \right] \rightarrow (\#)$$

which is the required transformed PDE.

Now for $R \rightarrow \infty$, we have from payoff that

$$H(R_T, T) = \left(1 - \frac{1}{T} R_T \right)^+$$

$$\Rightarrow \boxed{H(R_T, T) = 0 \text{ for } R_T \rightarrow \infty}$$

Now for PDE (#), when $R \rightarrow 0$, we have.

$$\left[\frac{\partial H}{\partial t} + \frac{1}{2} \sigma^2 R^2 \frac{\partial^2 H}{\partial R^2} + \frac{\partial H}{\partial R} = 0 \right] \rightarrow (*)$$

Claim: If H is bounded, then $R^2 \frac{\partial^2 H}{\partial R^2} \rightarrow 0$ as $R \rightarrow 0$

Proof: Assuming $R^2 \frac{\partial^2 H}{\partial R^2}$ is non-zero, we have.

$$\frac{\partial^2 H}{\partial R^2} = O\left(\frac{1}{R^2}\right)$$

Integrating this twice, we get.

$$H = O(\log R) + C_1 R + C_2.$$

$\Rightarrow H$ is not bounded for $R \rightarrow 0$

Hence this contradicts our assumption that H is bounded.

$\Rightarrow H$ is bounded

Hence we have

$$R^2 \frac{\partial^2 H}{\partial R^2} \rightarrow 0 \quad \text{as } R \rightarrow 0.$$

Using this in (*), we have the boundary condition at left boundary as follows:

$$\frac{\partial H}{\partial t} + \frac{\partial H}{\partial R} = 0 \quad \text{for } R \rightarrow 0$$

Hence final transformed PDE is as follows.

$$\frac{\partial H}{\partial t} + \frac{1}{2} \sigma^2 R^2 \frac{\partial^2 H}{\partial R^2} + (1 - rR) \frac{\partial H}{\partial R} = 0$$

$$H(R_T, T) = \left(1 - \frac{R_T}{T}\right)^+$$

$$H = 0 \quad \text{for } R \rightarrow \infty$$

$$\frac{\partial H}{\partial t} + \frac{\partial H}{\partial R} = 0 \quad \text{for } R = 0$$

(5) Strong-order of convergence

A discretization \hat{X} has strong order of convergence $\beta > 0$ if

$$E[\|\hat{X}(nh) - X(T)\|] \leq ch^\beta$$

for some constant c and sufficiently small h .

A discretization scheme has weak order of convergence β if

$$|E[f(\hat{X}(nh))] - E[f(X(T))]| \leq ch^\beta$$

for some constant c and all sufficiently small h .

and $\forall f \in C_p^{2\beta+2}$ where c may depend on f .

$C_p^{2\beta+2}$ consists of functions from \mathbb{R}^d to \mathbb{R} whose derivatives of order $0, 1, \dots, 2\beta+2$ are polynomially bounded. A function $g: \mathbb{R}^d \rightarrow \mathbb{R}$ is polynomially bounded if

$$|g(x)| \leq K(1 + \|x\|^q)$$

for some constants K and q and $\forall x \in \mathbb{R}^d$.

(6) Given SDE

$$\begin{cases} dx(t) = a(x(t)) dt + b(x(t)) dW(t) \\ x(0) = x_0 \end{cases}$$

We define the following operators:

$$\mathcal{L}^0 = a \frac{d}{dx} + \frac{1}{2} b^2 \frac{d^2}{dx^2}$$

$$\mathcal{L}^1 = b \frac{d}{dx}$$

For functions that depend explicitly on time t , we define \mathcal{L}^0 as follows:

$$\mathcal{L}^0 = \frac{\partial}{\partial t} + a \frac{\partial}{\partial x} + \frac{1}{2} b^2 \frac{\partial^2}{\partial x^2}$$

Hence for any twice differentiable function f , we have.

$$\mathcal{L}^0 f(x) = a(x) f'(x) + \frac{1}{2} b^2(x) f''(x)$$

and $\mathcal{L}^1 f(x) = b(x) f'(x)$

Now approximating the evolution of X over an interval $[t, t+h]$, we have

$$X(t+h) = X(t) + \int_t^{t+h} a(X(u)) du + \int_t^{t+h} b(X(u)) dW(u) \quad \text{--- (1)}$$

Now using Ito's formula on $a(x(u))$, we have

$$a(x(u)) = a(x(t)) + \int_t^u L^0 a(x(s)) ds + \int_t^u L^1 a(x(s)) dW(s) \quad (2)$$

Now using Euler approximation to each of the two integrals in above expression, we have

$$\begin{aligned} L^0 a(x(s)) &\approx L^0 a(x(t)) \\ \text{and } L^1 a(x(s)) &\approx L^1 a(x(t)) \end{aligned} \quad \left. \vphantom{\begin{aligned} L^0 a(x(s)) &\approx L^0 a(x(t)) \\ \text{and } L^1 a(x(s)) &\approx L^1 a(x(t)) \end{aligned}} \right\} \text{for } s \in [t, u]$$

Now putting these approximations in (2), we get.

$$a(x(u)) \approx a(x(t)) + L^0 a(x(t)) \int_t^u ds + L^1 a(x(t)) \int_t^u dW(s) \quad (3)$$

Now putting this approximation in (1) we get approximation of first integral

$$\begin{aligned} \int_t^{t+h} a(x(u)) du &\approx a(x(t)) h + L^0 a(x(t)) \int_t^{t+h} \int_t^u ds du \\ &\quad + L^1 a(x(t)) \int_t^{t+h} \int_t^u dW(s) du \end{aligned}$$

$$\int_t^{t+h} a(x(u)) du \equiv a(x(t)) h + L^0 a(x(t)) I_{(0,0)} + L^1 a(x(t)) I_{(1,0)} \quad (4)$$

where $I_{(0,0)} = \int_t^{t+h} \int_t^u ds du$ and $I_{(1,0)} = \int_t^{t+h} \int_t^u dW(s) du$
are double integrals

Now for approximating the second integral in (1), we have for approximation of integrand $b(x(u))$ for $u \in [t, t+h]$ as follows:

$$b(x(u)) = b(x(t)) + \int_t^u \mathcal{L}^0 b(x(s)) ds + \int_t^u \mathcal{L}' b(x(s)) dW(s)$$

$$b(x(u)) \approx b(x(t)) + \mathcal{L}^0 b(x(t)) \int_t^u ds + \mathcal{L}' b(x(t)) \int_t^u dW(s) \quad (5)$$

where Euler approximation is used for approximating the two integrands involved i.e.

$$\mathcal{L}^0 b(x(s)) \approx \mathcal{L}^0 b(x(t)) \quad \text{and} \quad \mathcal{L}' b(x(s)) \approx \mathcal{L}' b(x(t))$$

for $s \in [t, u]$.

Then the second integral in (1) is approximated as:

$$\begin{aligned} \int_t^{t+h} b(x(u)) dW(u) &\approx b(x(t)) [W(t+h) - W(t)] \\ &+ \mathcal{L}^0 b(x(t)) \int_t^{t+h} \int_t^u ds dW(u) + \mathcal{L}' b(x(t)) \int_t^{t+h} \int_t^u dW(s) dW(u) \end{aligned}$$

$$\Rightarrow \int_t^{t+h} b(x(u)) dW(u) \approx b(x(t)) [W(t+h) - W(t)] + \mathcal{L}^0 b(x(t)) I_{(0,1)} + \mathcal{L}' b(x(t)) I_{(1,1)} \quad (6)$$

where $I_{(0,1)} = \int_t^{t+h} \int_t^u ds dW(u)$ and $I_{(1,1)} = \int_t^{t+h} \int_t^u dW(s) dW(u)$

Now Calculating the double integrals involved as follows:

$$* I_{(0,0)} = \int_t^{t+h} \int_t^u ds du = \int_t^{t+h} (u-t) du = \frac{u^2}{2} \Big|_t^{t+h} - ht$$

$$\Rightarrow \boxed{I_{(0,0)} = \frac{h^2}{2}}$$

$$* I_{(1,1)} = \int_t^{t+h} [W(u) - W(t)] dW(u) = \int_t^{t+h} W(u) dW(u) - W(t) \int_t^{t+h} dW(u)$$

$$= \frac{1}{2} [(\Delta W)^2 - h] \quad \text{where } \Delta W = W(t+h) - W(t)$$

Here taking $f(x) = x^2$ and using Ito's formula, we have

$$W^2(t+h) - W^2(t) = 2 \int_t^{t+h} W(u) dW(u) + \frac{1}{2} \cdot 2 \int_t^{t+h} dt$$

$$\Rightarrow \int_t^{t+h} W(u) dW(u) = \frac{1}{2} [W^2(t+h) - W^2(t) - h]$$

Substituting this and rearranging we will get req. form.

$$* I_{(0,1)} = \int_t^{t+h} \int_t^u ds dW(u) = \int_t^{t+h} (u-t) dW(u)$$

$$= (u-t)W(u) \Big|_t^{t+h} - \int_t^{t+h} 1 \cdot dW(u)$$

Using Integrating by parts or applying Ito's lemma to $t \cdot W(t)$

$$= h W(t+h) - \int_t^{t+h} W(u) du$$

$$= h [W(t+h) - W(t)] - \int_t^{t+h} [W(u) - W(t)] du$$

$$\Rightarrow \boxed{I_{(0,1)} = h \Delta W - I_{(1,0)}}$$

$$* I_{(1,0)} = \int_t^{t+h} [W(u) - W(t)] du$$

Now given $W(t)$, $I_{(1,0)}$ and increment $\Delta W = W(t+h) - W(t)$ are jointly Normal.

Now conditional mean of both ΔW and $I_{(1,0)}$ is 0.

Conditional variance of $\Delta W = W(t+h) - W(t)$ is $t+h-t = h$.

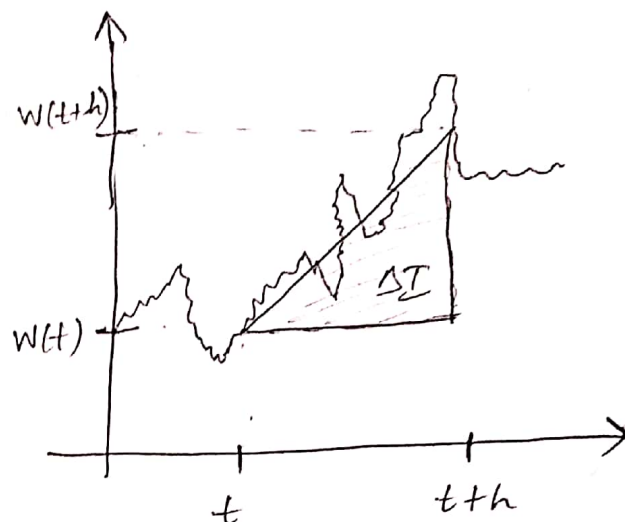
Conditional variance of $I_{(1,0)} \neq h^3/2$ { $\begin{matrix} \text{beoz} \\ W(t+h) - W(t) \sim N(0, h) \end{matrix} \}$

$$\left\{ \begin{aligned} \text{beoz } \text{Var} \left[\int_0^T W(u) du \right] &= 2 \int_0^T \int_0^s \text{Cov}[W(u), W(s)] du ds \\ &= 2 \int_0^T \int_0^s u du ds = \int_0^T s^2 ds = T^3/3 \end{aligned} \right\}$$

Now we have that

$$E[I_{(1,0)} | W(t), \Delta W] = \frac{h}{2} \Delta W.$$

In the plot, shaded area is ΔI
Hence given $W(t)$ and $W(t+h)$,
the conditional expectation of W
at any intermediate time lies on
straight line connecting these end-points.



\Rightarrow conditional expectation of ΔI = Area of triangle
 $= \frac{1}{2} h \Delta W$

So, we have $E[I_{(1,0)} \Delta W] = h^2/2$

Hence we can simulate $\Delta W = W(t+h) - W(t)$ and $I_{(1,0)}$ as

$$\begin{pmatrix} \Delta W \\ \Delta I \end{pmatrix} \sim N\left(0, \begin{pmatrix} h & h^2/2 \\ h^2/2 & h^3/3 \end{pmatrix}\right)$$

Now expanding L^0 and L^1 explicitly we have.

$$L^0 a = aa' + \frac{1}{2} b^2 a'' ; L^1 a = ba'$$

$$L^0 b = ab' + \frac{1}{2} b^2 b'' ; L^1 b = bb'$$

Then putting in (4), we have

$$\int_t^{t+h} a(x(u)) du \equiv ah + \left(aa' + \frac{1}{2} b^2 a''\right) I_{(0,0)} + ba' (I_{(1,0)})$$

Putting in (6), we have.

$$\int_t^{t+h} b(x(u)) dW(u) \equiv b \Delta W + \left(ab' + \frac{1}{2} b^2 b''\right) I_{(0,1)} + bb' I_{(1,1)}$$

Then putting in Eqn (1), we have

$$X(t+h) \approx X(t) + ah + b\Delta W + \left(aa' + \frac{1}{2}b^2a''\right)I_{(0,0)} \\ + \left(ab' + \frac{1}{2}b^2b''\right)I_{(0,1)} + ba'I_{(1,0)} + bb'I_{(1,1)}$$

with function a, b and their derivatives being evaluated at $X(t)$.

Finally using approximation of all double-integrals in above approximation, the 2nd order scheme is as follows:

$$\hat{X}((i+1)h) = \hat{X}(ih) + ah + b\Delta W + \left(ab' + \frac{1}{2}b^2b''\right)(\Delta W\Delta t - \Delta I) \\ + ab'\Delta I + \frac{1}{2}bb'[\Delta W^2 - h] + \left(aa' + \frac{1}{2}b^2a''\right)\frac{h^2}{2}$$

with function a, b and their derivatives being evaluated at $\hat{X}(ih)$. This is called Milstein second order scheme.