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MA-473 : COMPUTATIONAL FINANCE

END SEM → PAPER-I

① Given PDE

$$\left\{ \begin{array}{l} \frac{\partial V}{\partial t} + \frac{1}{2} \sigma^2 S^2 \frac{\partial^2 V}{\partial S^2} + (\gamma - \delta) S \frac{\partial V}{\partial S} - \gamma V = 0, \quad 0 \leq S, \\ t \leq T \\ V(S, T) = V_T(S), \quad 0 \leq S \end{array} \right. \quad \rightarrow (1)$$

Using following transformation

$$\left\{ \begin{array}{l} \xi = \frac{S}{S + P_m} \\ \tau = T - t \\ V(S, t) = (S + P_m) \bar{V}(\xi, \tau) \end{array} \right.$$

then we have $\boxed{S = \frac{P_m \xi}{1 - \xi}} \quad \text{and} \quad \boxed{S + P_m = \frac{P_m}{1 - \xi}} \quad (2) \quad (3)$

Now $\frac{d\xi}{dS} = 1 \cdot \frac{1}{S + P_m} + S \cdot \left(\frac{-1}{(S + P_m)^2} \right) \cdot 1$

$$\Rightarrow \frac{d\xi}{dS} = \frac{P_m}{(S + P_m)^2}$$

$$\Rightarrow \boxed{\frac{d\xi}{dS} = \frac{(1 - \xi)^2}{P_m}} \rightarrow (4) \quad \left\{ \text{Using (3)} \right\}$$

Now we have

$$\frac{\partial V}{\partial t} = \frac{\partial}{\partial t} \left[(S+P_m) \bar{V}(\xi, \tau) \right] = (S+P_m) \frac{\partial \bar{V}(\xi, \tau)}{\partial t}$$

$$= (S+P_m) \frac{\partial \bar{V}}{\partial \tau} \cdot \frac{d\tau}{dt} \Rightarrow \boxed{\frac{\partial V}{\partial t} = -(S+P_m) \frac{\partial \bar{V}}{\partial \tau}}$$

Using (3), we have

$$\boxed{\frac{\partial V}{\partial t} = -\frac{P_m}{1-\xi} \frac{\partial \bar{V}}{\partial \tau}}$$

$$\text{Now } \frac{\partial V}{\partial S} = \frac{\partial}{\partial S} \left[(S+P_m) \bar{V}(\xi, \tau) \right] = 1 \cdot \bar{V}(\xi, \tau) + (S+P_m) \frac{\partial \bar{V}(\xi, \tau)}{\partial S}$$

$$= \bar{V} + (S+P_m) \frac{\partial \bar{V}}{\partial \xi} \cdot \frac{d\xi}{dS}$$

$$= \bar{V} + \frac{P_m}{1-\xi} \cdot \frac{\partial \bar{V}}{\partial \xi} \cdot \frac{(1-\xi)^2}{P_m} \quad \left\{ \begin{array}{l} \text{Using (3)} \\ \text{and (4)} \end{array} \right\}$$

$$\Rightarrow \boxed{\frac{\partial V}{\partial S} = \bar{V} + (1-\xi) \frac{\partial \bar{V}}{\partial \xi}}$$

$$\text{Now } \frac{\partial^2 V}{\partial S^2} = \frac{\partial}{\partial S} \left[\frac{\partial V}{\partial S} \right] = \frac{\partial}{\partial S} \left[\bar{V} + (1-\xi) \frac{\partial \bar{V}}{\partial \xi} \right]$$

$$= \frac{\partial \bar{V}}{\partial \xi} \cdot \frac{d\xi}{dS} + \frac{\partial}{\partial S} \left[(1-\xi) \cdot \frac{\partial \bar{V}}{\partial \xi} \right]$$

$$\begin{aligned} \Rightarrow \frac{\partial^2 \bar{V}}{\partial s^2} &= \frac{\partial \bar{V}}{\partial \xi_1} \cdot \frac{d\xi_1}{ds} + \frac{\partial}{\partial \xi_1} \left[(1-\xi_1) \frac{\partial \bar{V}}{\partial \xi_1} \right] \cdot \frac{d\xi_1}{ds} \\ &= \frac{\partial \bar{V}}{\partial \xi_1} \cdot \frac{d\xi_1}{ds} - \frac{\partial \bar{V}}{\partial \xi_1} \cdot \frac{d\xi_1}{ds} + (1-\xi_1) \cdot \frac{\partial^2 \bar{V}}{\partial \xi_1^2} \cdot \frac{d\xi_1}{ds} \end{aligned}$$

$$\Rightarrow \boxed{\frac{\partial^2 \bar{V}}{\partial s^2} = \frac{(1-\xi_1)^3}{P_m} \frac{\partial^2 \bar{V}}{\partial \xi_1^2}}$$

Now let $\bar{T}(\xi_1) = \sigma(S(\xi_1)) = \sigma\left(\frac{P_m \xi_1}{1-\xi_1}\right)$

Now putting all values in PDE (1), we get

$$\begin{aligned} \frac{\partial \bar{V}}{\partial \tau} &= \frac{\bar{T}^2(\xi_1) \cdot \xi_1^2 (1-\xi_1)^2}{2} \cdot \frac{\partial^2 \bar{V}}{\partial \xi_1^2} + (r-s) \xi_1 (1-\xi_1) \frac{\partial \bar{V}}{\partial \xi_1} \\ &\quad - [\tau(1-\xi_1) + s\xi_1] \bar{V} \end{aligned}$$

$0 \leq \xi_1 < 1, \tau \geq 0$

Now assuming \bar{V} is smooth function of ξ_1 .

then Eqⁿ also holds at $\xi_1 = 1$.

Now $V(s, \tau) = (s + P_m) \bar{V}(\xi_1, 0) = \bar{V}(\xi_1, 0) \frac{P_m}{1-\xi_1}$

Hence the condition $V(S,T) = V_T(S)$ can be re-written as:

$$\bar{V}(\xi, 0) = V_T\left(\frac{P_m \xi}{1 - \xi}\right) \cdot \frac{1 - \xi}{P_m}$$

This is ~~not~~ Initial condition for PDE (5).

Boundary Conditions

At $\xi = 0$ and $\xi = 1$, PDE (5) degenerates to O.D.E at boundaries

$$\boxed{\text{At } \xi = 0} \quad \frac{\partial \bar{V}(0, \tau)}{\partial \tau} = -r \bar{V}(0, \tau)$$

$$\Rightarrow \frac{d\bar{V}(0, \tau)}{d\tau} = -r d\tau$$

$$\Rightarrow \boxed{\bar{V}(0, \tau) = \bar{V}(0, 0) e^{-r\tau}}$$

Similarly $\boxed{\text{at } \xi = 1}$

$$\frac{\partial \bar{V}(1, \tau)}{\partial \tau} = -s \bar{V}(1, \tau)$$

$$\Rightarrow \frac{d\bar{V}(1, \tau)}{\bar{V}(1, \tau)} = -s d\tau$$

$$\Rightarrow \boxed{\bar{V}(1, \tau) = \bar{V}(1, 0) e^{-s\tau}}$$

Hence for PDE (5), the two solution of ODE provide boundary values and no. B.C. are needed for the PDE to have unique solution.

Hence the final PDE becomes:

$$\left\{ \begin{aligned} \frac{\partial \bar{V}}{\partial \tau} &= \frac{1}{2} \sigma^2(q) q^2 (1-q)^2 \frac{\partial^2 \bar{V}}{\partial q^2} + (r-s)q(1-q) \frac{\partial \bar{V}}{\partial q} - [r(1-q) + sq] \bar{V} \\ 0 &\leq q \leq 1, 0 \leq \tau \end{aligned} \right.$$

$$\bar{V}(q, 0) = \frac{1-q}{p_m} V_T \left(\frac{p_m q}{1-q} \right), 0 \leq q \leq 1 \rightarrow \text{I.C.}$$

$$\bar{V}(0, \tau) = \bar{V}(0, 0) e^{-r\tau}$$

$$\bar{V}(1, \tau) = \bar{V}(1, 0) e^{-s\tau}$$

} B.C.

Hence B.S. PDE is transformed from infinite domain $s \geq 0$ to finite domain $0 \leq q \leq 1$.

(2)

American option pricing problem can be formulated as variational problem with class of comparison functions defined as:

$$K = \{ v \in C^0 : \frac{\partial v}{\partial \tau} \text{ is piecewise } C^0, \$$

$$v(x, \tau) \geq g(x, \tau) \quad \forall x, \tau.$$

$$v(x, 0) = g(x, 0)$$

$$v(x_{\max}, \tau) = g(x_{\max}, \tau)$$

$$v(x_{\min}, \tau) = g(x_{\min}, \tau) \quad \} \quad y$$

where

$$g(x, \tau) = \begin{cases} \exp\left\{\frac{\tau}{4}((q_\delta-1)^2 + 4q)\right\} \max\left\{e^{\frac{x}{2}(q_\delta-1)} - e^{\frac{x}{2}(q_\delta+1)}, 0\right\} & \text{for put} \\ \exp\left\{\frac{\tau}{4}((q_\delta-1)^2 + 4q)\right\} \max\left\{e^{\frac{x}{2}(q_\delta+1)} - e^{\frac{x}{2}(q_\delta-1)}, 0\right\} & \text{for call} \end{cases}$$

$$\text{with } q = \frac{2\sigma}{\sigma^2} \text{ and } q_\delta = \frac{2(\delta-\delta)}{\sigma^2}$$

Now, let y be the exact solution of American option pricing problem formulated as Linear Complementarity Problem. As solution of Partial Differential Inequality, y is C^2 -smooth on continuation region and $y \in K$.

Also for $v \in K$

$$v \geq g, \quad \frac{\partial y}{\partial \tau} - \frac{\partial^2 y}{\partial x^2} \geq 0$$

$$\Rightarrow \int_{x_{\min}}^{x_{\max}} \left(\frac{\partial y}{\partial \tau} - \frac{\partial^2 y}{\partial x^2} \right) (v - g) dx \geq 0 \quad - (1)$$

Also from the complementarity condition, we have.

$$\int_{x_{\min}}^{x_{\max}} \left(\frac{\partial y}{\partial \tau} - \frac{\partial^2 y}{\partial x^2} \right) (y - g) dx = 0 \quad - (2)$$

Now Subtracting (2) from (1), we get

$$\int_{x_{\min}}^{x_{\max}} \left(\frac{\partial y}{\partial \tau} - \frac{\partial^2 y}{\partial x^2} \right) (v - y) dx \geq 0$$

$$\Rightarrow \int_{x_{\min}}^{x_{\max}} \frac{\partial y}{\partial \tau} (v - y) dx - \int_{x_{\min}}^{x_{\max}} (v - y) \frac{\partial^2 y}{\partial x^2} dx \geq 0$$

$$\Rightarrow \int_{x_{\min}}^{x_{\max}} \frac{\partial y}{\partial \tau} (v - y) dx - \left[(v - y) \frac{\partial y}{\partial x} \right]_{x_{\min}}^{x_{\max}} - \int_{x_{\min}}^{x_{\max}} \left(\frac{\partial v}{\partial x} - \frac{\partial y}{\partial x} \right) \frac{\partial y}{\partial x} dx \geq 0$$

Now at boundary points x_{\min} and x_{\max} , we have

Using Integration by Parts \downarrow (3)
 in 2nd Integral
 1st func. = $v - y$, 2nd func. = $\frac{\partial^2 y}{\partial x^2}$.

and

$$y(x_{\min}, \tau) = g(x_{\min}, \tau) \\ \text{and } y(x_{\max}, \tau) = g(x_{\max}, \tau)$$

$$v(x_{\min}, \tau) = g(x_{\min}, \tau) \\ \text{and } v(x_{\max}, \tau) = g(x_{\max}, \tau)$$

Hence, we have

$$v(x_{\min}, \tau) = y(x_{\min}, \tau) \quad \text{and} \quad v(x_{\max}, \tau) = y(x_{\max}, \tau)$$

Hence, the term
$$\left[\frac{\partial y}{\partial x} (v - y) \right]_{x_{\min}}^{x_{\max}} = 0$$

Putting this in eqⁿ (3), we get

→ (4)

$$I(y; v) = \int_{x_{\min}}^{x_{\max}} \left(\frac{\partial y}{\partial \tau} (v - y) + \frac{\partial y}{\partial x} \left(\frac{\partial v}{\partial x} - \frac{\partial y}{\partial x} \right) \right) dx \geq 0 \quad \forall v \in K$$

Now, this inequality holds $\forall v \in K$. Hence choosing $v = y$ we have $I(y; y) = 0$ and the integral $I(y; v)$ takes its minimum value i.e.

$$\min_{v \in K} I(y; v) = I(y; y) = 0$$

(2)
(a) We now consider $\hat{y} \in K$ that is not C^2 -smooth on continuation region. We want to find $\hat{y} \in K$

$$\begin{aligned} & \text{s.t. } I(\hat{y}; v) \geq 0 \quad \forall v \in K \\ & \text{and } \inf_{v \in K} I(\hat{y}; v) = 0 \end{aligned} \quad \left. \vphantom{\begin{aligned} & \text{s.t. } I(\hat{y}; v) \geq 0 \quad \forall v \in K \\ & \text{and } \inf_{v \in K} I(\hat{y}; v) = 0 \end{aligned}} \right\} \text{ This is called Weak Formulation}$$

Now, consider the following approximations for \hat{y} and v :

$$\begin{aligned} \hat{y} &= \sum_i w_i(\tau) \phi_i(x) \\ v &= \sum_i v_i(\tau) \phi_i(x) \end{aligned} \quad \left. \vphantom{\begin{aligned} \hat{y} &= \sum_i w_i(\tau) \phi_i(x) \\ v &= \sum_i v_i(\tau) \phi_i(x) \end{aligned}} \right\} \text{ where } \phi_i \text{ are the basis functions.}$$

This formulation separates the independent variables τ and x . Hence the same x -grid is applied for all τ , which

results in rectangular grid in (x, τ) -plane. The time dependence is incorporated in coeff. functions w_i and v_i . Since ϕ_i represent x_i -grid, we

consider semi-discretization of eqn (4) as follows:

$$\int \left\{ \left(\sum_i \frac{dw_i}{d\tau} \phi_i \right) \left(\sum_j (v_j - w_j) \phi_j \right) + \left(\sum_i w_i \phi_i' \right) \left(\sum_j (v_j - w_j) \phi_j' \right) \right\} dx \geq 0$$

$$\Rightarrow \left[\sum_i \sum_j \frac{dw_i}{d\tau} (v_j - w_j) \int \phi_i \phi_j dx + \sum_i \sum_j w_i (v_j - w_j) \int \phi_i' \phi_j' dx \geq 0 \right]$$

Now writing the inequality into vector-matrix form, we have

$$\left(\frac{dw}{dt}\right)^T B(v-w) + w^T A(v-w) \geq 0$$

$$\Rightarrow \boxed{(v-w)^T \left(B \frac{dw}{dt} + Aw \right) \geq 0} \rightarrow (\#)$$

where matrices A and B are constructed using assembling algorithm s.t.

$$a_{ij} = \int_{\mathcal{D}} \phi_i \phi_j dx \quad \text{and} \quad b_{ij} = \int_{\mathcal{D}} \phi_i' \phi_j' dx$$

where \mathcal{D} is domain.

For equidistant steps, we have for ϕ_i as hat functions

$$A = \frac{1}{h} \begin{pmatrix} 1 & -1 & & & \\ -1 & 2 & -1 & & \\ & -1 & 2 & \ddots & \\ & & & \ddots & \ddots \\ & & & & 2 & -1 \\ & & & & -1 & 1 \end{pmatrix}$$

$$B = \frac{h}{6} \begin{pmatrix} 2 & 1 & & & \\ 1 & 4 & 1 & & \\ & 1 & 4 & \ddots & \\ & & & \ddots & \ddots \\ & & & & 4 & 1 \\ & & & & 1 & 2 \end{pmatrix}$$

Now time is discretized.

Define $\omega^{(n)} = \omega(\tau_n)$ and $v^{(n)} = v(\tau_n)$

Now considering

$$\frac{w_{i,n+1} - w_{i,n}}{\Delta \tau} = \theta \left(\frac{w_{i+1,n+1} - 2w_{i,n+1} + w_{i-1,n+1}}{\Delta x^2} \right) + (1-\theta) \left(\frac{w_{i+1,n} - 2w_{i,n} + w_{i-1,n}}{\Delta x^2} \right)$$

Now using this approximation in (#), we get

$$\left(v^{(n+1)} - w^{(n+1)} \right)^T \left(B \frac{1}{\Delta \tau} (w^{(n+1)} - w^{(n)}) + \theta A w^{(n+1)} + (1-\theta) A w^{(n)} \right) \geq 0 \quad \forall n.$$

Now for $\theta = \frac{1}{2}$, we have Crank-Nicolson method.

$$\Rightarrow \left(v^{(n+1)} - w^{(n+1)} \right)^T \left((B + \Delta \tau \theta A) w^{(n+1)} + (\Delta \tau (1-\theta) A - B) w^{(n)} \right) \geq 0$$

Now let $C = B + \Delta \tau \theta A$

$$r = (B - \Delta \tau (1-\theta) A) w^{(n)}$$

then we have.

$$\left(v^{(n+1)} - w^{(n+1)} \right)^T (C w^{(n+1)} - r) \geq 0$$

which is fully discretized version of $\mathcal{L}(\hat{y}; v) \geq 0$

Side conditions

We have $\hat{y}(x, \tau) \geq g(x, \tau)$

$$\Rightarrow \sum w_i(\tau) \phi_i(x) \geq g(x, \tau)$$

For hat functions ϕ_i $\left(\begin{array}{l} \phi_i(x_i) = 1 \\ \text{and } \phi_i(x_j) = 0 \text{ for } j \neq i \end{array} \right)$
and $x = x_j$, we

have $w_j(\tau) \geq g(x_j, \tau)$.

Then with $\tau = \tau_n$, we have

$$w^{(n)} \geq g^{(n)}$$

and similarly $v^{(n)} \geq g^{(n)}$

(3)

For showing equivalence of FEM and FDM, consider FEM and FDM version American option pricing problem as follows:

FEM

Construct w s.t.

$$\forall v \geq g.$$

$$(v-w)^T (cw-r) \geq 0, w \geq g$$

FDM

Find w s.t.

$$cw-r \geq 0, w \geq g$$

$$(cw-r)^T (w-g) = 0$$

The proof of equivalence of FDM and FEM is as follows:

$$(FDM) \Rightarrow (FEM)$$

Let w solve FDM, so $w \geq g$ and

$$(v-w)^T (cw-r) = (v-g)^T (cw-r) - (w-g)^T (cw-r)$$

Now, $cw-r \geq 0$ and $(w-g)^T (cw-r) = 0$

$$\Rightarrow (v-w)^T (cw-r) = (v-g)^T (cw-r) \geq 0, \forall v \geq g.$$

$$\Rightarrow \boxed{(v-w)^T (cw-r) \geq 0, \forall v \geq g} \quad \text{hence } w \text{ solves FEM.}$$

$$(FEM) \Rightarrow (FDM)$$

Let w solve FEM, so $w \geq g$ s.t.

$$v^T(cw-r) \geq w^T(cw-r) \quad \forall v \in K$$

Now suppose k^{th} component of $cw-r$ is negative and make v_k arbitrarily large. Then LHS becomes arbitrarily small, which is contradiction.

\Downarrow

$$cw-r \geq 0.$$

$$\text{Hence } w \geq g \Rightarrow \boxed{(w-g)^T(cw-r) \geq 0} \rightarrow (*)$$

Now in (FEM), we have

$$(v-w)^T(cw-r) \geq 0$$

$$\text{Put } v=g, \text{ then } \boxed{(w-g)^T(cw-r) \leq 0} \rightarrow (\#)$$

Using (*) and (#), we conclude that

$$\boxed{(w-g)^T(cw-r) = 0}$$

Hence w solves FDM.

Combining both parts, we have FEM and FDM formulations are equivalent i.e. solution of one problem satisfies another as well.