

So) (3): using FTCS.

$$w_{i,v+1} = \lambda w_{i-1,v} + (1-2\lambda) w_{i,v} + \lambda w_{i+1,v}$$

$$\lambda = \frac{\Delta t}{\Delta x^2}$$

So coeff matrix =

$$A = \begin{bmatrix} 1-2\lambda & \lambda & 0 & \dots & 0 \\ \lambda & 1-2\lambda & \lambda & \dots & 0 \\ 0 & \lambda & 1-2\lambda & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & \dots & 0 & \lambda & 1-2\lambda \end{bmatrix}$$

$$w^{v+1} = A w^v \quad \text{for } v = 0, 1, 2, \dots, v_{\max}-1$$

$$\text{Error analysis: } w^{v+1} = A w^v + d^v$$

$$w^v \rightarrow \text{exact vectors of } w^{v+1} = A w^v + d^v$$

$w^v \rightarrow$  subject to rounding errors.

$$\text{less errors } e^v = \bar{w}^{(v)} - w^v$$

and result is

$$\bar{w}^{(v+1)} = A \bar{w}^v + \bar{d}^v \quad \text{rounding error}$$

Effect of initial rounding error  $e^{(0)}$  have on the iteration is (let  $\delta^v = 0$  for  $v > 1$ )

$$A e^v = A \bar{w}^v - A w^v = \bar{w}^{(v+1)} - w^{v+1} = e^{v+1}$$

$$\Rightarrow A e^v = e^{v+1} \Rightarrow \boxed{e^v = A^v e^{(0)}}$$

for stability we require  $A^v e^{(0)} \rightarrow 0$  for  $v \rightarrow \infty$

hence  $\lim_{v \rightarrow \infty} A^v e^{(0)} = 0$

which is true iff spectral radius of  $A < 1$

$$\rho(A) = \max |\mu_i| < 1$$

where  $\mu_1, \dots, \mu_{m-1} \rightarrow$  eigen values of  $A$

for  $G = \begin{bmatrix} \alpha & \beta & \dots & 0 \\ \gamma & & & \\ \vdots & & & \beta \\ \vdots & & \gamma & \alpha \end{bmatrix} \in \mathbb{R}^{N \times N}$

the eigen values are  $\mu_k^G = \alpha + 2\beta \sqrt{\lambda/\beta} \cos\left(\frac{k\pi}{N+1}\right)$

$\therefore A = I - \lambda \begin{bmatrix} 2 & 1 & \dots & 0 \\ 1 & 2 & \dots & 1 \\ \vdots & \vdots & \ddots & \vdots \\ 1 & \dots & 1 & 2 \end{bmatrix}$   
 $= I - \lambda G$

$\mu_k^A = 1 - \mu_k^G$       $\alpha = 2, \beta = \gamma = -1, N = m-1$

$$\mu_k^G = 2 - 2\cos\left(\frac{k\pi}{m}\right) = 4\sin^2\left(\frac{k\pi}{2m}\right)$$

$$\mu_k^A = \left| 1 - 4\lambda \sin^2\left(\frac{k\pi}{2m}\right) \right|$$

$\Rightarrow$  for stability  $\mu_k^A < 1 \Rightarrow \left| 1 - 4\lambda \sin^2\left(\frac{k\pi}{2m}\right) \right| < 1$   
 $k = 1, 2, \dots, m-1$

$\Rightarrow \lambda > 0$  and  $-1 < 1 - 4\lambda \sin^2\left(\frac{k\pi}{2m}\right) < 1$

$\Rightarrow \lambda \sin^2\left(\frac{k\pi}{2m}\right) < \frac{1}{2} \Rightarrow \lambda < \frac{1}{2}$



Naman Goyal

classmate

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Hence  $0 < \lambda \leq 1/2$  is sufficient.

Hence for above  $\lambda$ , the explicit method FTCS

$w^{n+1} = w^n$  is stable if  $\lambda \leq \frac{\Delta t}{\Delta x^2}$ .

$$1 \geq (\Delta t / \Delta x^2) = (\lambda) \Delta t$$

Time step  $\Delta t \leq \frac{1}{\lambda}$  is required.

Q.4) To show:   
 primal problem:

$$\text{for } \hat{b} = b - Ag$$

$$Ax - y = \hat{b} \quad y \geq 0, \quad x^T y = 0$$

is equivalent to

$$\left( \min_{x \geq 0} G(x) \right) \quad G(x) = \frac{1}{2} (x^T A x) - b^T x$$

is strictly convex.

Sol<sup>n</sup> (4)

↳ Since the derivatives of  $G$  are  $G_x = Ax - \hat{b}$  and  $G_{xx} = A$ . we can use lemma and can see that  $A$  has +ve eigen values. Hence the Hessian matrix  $G_{xx}$  is symmetric and positive definite. So  $G$  is strictly convex and has a unique minimum on each convex set in  $R^n$  for  $x \geq 0$ .

eg. 2.20 The Kuhn and Tucker theorem minimizes  $G(x)$  under  $H_i(x) \leq 0 \quad i=1, \dots, m$ .

Acc to the theorem, a vector  $x_0$  to be a minimum is equivalent to existence of a Lagrange multiplier  $y \geq 0$  with

$$\text{grad } G(x_0) + \left( \frac{\partial H(x_0)}{\partial x} \right)^T y = 0 \quad y^T H(x_0) = 0$$

The set  $\{x \geq 0\}$  leads to define  $H(x) = -x$ . Hence putting this in above eq<sup>n</sup> leads to cond<sup>n</sup>  $Ax - \hat{b} + (-I)^T y = 0, \quad y^T x = 0$  which is the initial primal problem.



Naman Goyal  
180123029

classmate  
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Q5: early exercise where  $(S_f(t))$  ( $t < T$ )

$$S_f(t) < \lim_{t \rightarrow T} S_f(t) = \min(K, r/s K)$$

for  $t < T$ : the value of  $V_P^{\text{American}} = \text{payoff}$

sol<sup>n</sup> (5)  $\Rightarrow V_P^{\text{Amer.}}(S, T) = K - S$  for  $S < K$ .

But this is Black Scholes eq<sup>n</sup>:

$$\frac{\partial V(S, t)}{\partial t} + 0 - (r - \delta)S - rV = 0 \Rightarrow \frac{\partial V}{\partial t} = r(S + V) - \delta S$$

$$\Rightarrow \frac{\partial V(S, T)}{\partial t} = (rK - \delta S)$$

$$(\text{claim}) \Rightarrow rK - \delta S \leq 0 \Rightarrow \frac{\partial V(S, T)}{\partial t} \leq 0 \quad \text{--- (1)}$$

otherwise for  $t \rightarrow T \Rightarrow V \geq \text{payoff}$  which is not possible. Hence for  $t = T$  and  $S < K$ .

$$rK - \delta S \leq 0 \Rightarrow \boxed{S \geq r/s K}$$

Hence, we can also deduce that  $\delta > r$  otherwise  $S > K$  which is not good.

$\hookrightarrow$  There could exist multiple possibilities like either

$$S_f(T) = \lim_{t \rightarrow T} S_f(t) \quad \text{satisfies } S_f(T) = r/s K$$

or other 2 cases possible maybe

$$(i) S_f(T) < r/s K \quad (ii) S_f(T) > r/s K$$



(i) There is  $S$  such that  $S_f(T) < S < \frac{r}{\delta} K$ .

Then  $\frac{\partial v(S, T)}{\partial t} = rK - \delta S > 0$

but claim (i) this is not possible.

(i') There is  $S$  such that  $\frac{r}{\delta} K < S < S_f(T)$ .

Then  $rK < \delta S$  and  $K(e^{rdt} - 1) < S(e^{\delta dt} - 1)$

which implies that dividend earns more than the interest on  $K$ , early exercise ~~should be~~ shouldn't be executed. Thus it contradicts the fact that  $S < S_f(T)$ .

last case  $\rightarrow \delta \leq r$ .

$\therefore S_f(T) > K$  is not possible.  $\Rightarrow$

Assume  $S_f(T) < K$  then for  $S_f(T) < S < K$  and  $t \approx T$ .

$\frac{\partial v}{\partial t} = rK - \delta S$  but  $rK - \delta S > 0$

has to be  $\leq 0$

which can't be possible as LHS  $\leq 0$  and RHS  $> 0$ .

Hence,  $S_f(T) = K$  for  $\delta \leq r$ .

Hence we can conclude from above discussion that:

$S_f(t) = \lim_{t \rightarrow T} S_f(t) = \min(K, \frac{r}{\delta} K)$ .