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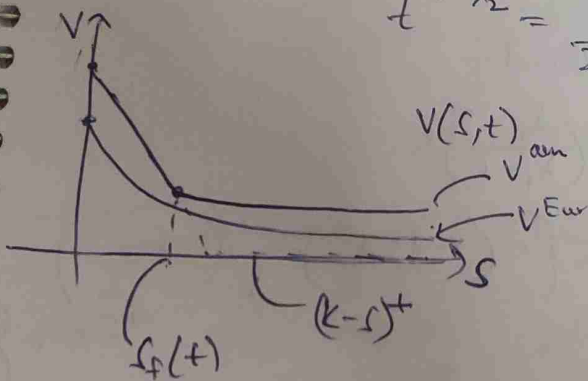
$$B\dot{w} + b = -Aw - a$$

$$w_j(0) = \alpha(x_j) - \phi_0(x_j, 0)$$

Crank-Nicolson:-

$$\left(B + \frac{\delta t A}{2}\right) w^{n+1} = \left(B - \frac{\delta t A}{2}\right) w^{(n)} - \frac{\delta t}{2} \left(a^{(n)} + a^{(n+1)} + b^{(n)} + b^{(n+1)}\right)$$

$$t^{n+1/2} = \frac{1}{2} (t^n + t^{(n+1)})$$

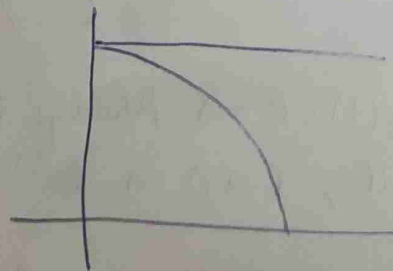
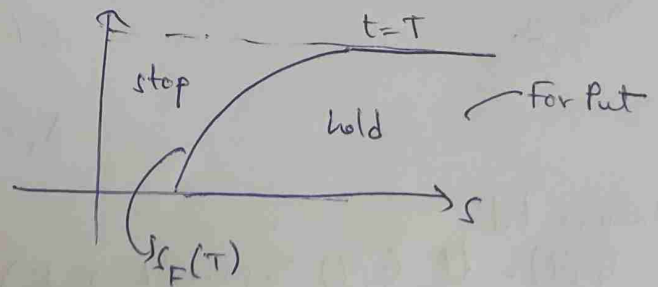


$$V_p^{Am}(s, t) \geq (k-s)^+$$

$$V_c^{Am}(s, t) \geq (s-k)^+$$

$$V_p(s, t) = k - s$$

$$V_p^{Am}(s, t) > (k-s)^+, \quad s > s_f(t)$$



The  $f(t)$  divides the curve into 2 parts —

The Early exercise is monotonic, smooth.

$f(t)$  is not known a priori. American opt<sup>n</sup> is referred as free bday problem.

Early - Exercise Curve:-

Put:-

1)  $f_t$  is cont diff  $0 \leq t < T$

2)  $f_t$  is non decreasing

3)  $f_t(t) > \left(\frac{\lambda_2}{\lambda_2 - 1}\right) K$

$$\lambda_2 = \frac{1}{\sigma^2} \left\{ -\left(r - \delta - \frac{\sigma^2}{2}\right) - \sqrt{\left(r - \delta - \frac{\sigma^2}{2}\right)^2 + 2\sigma^2 r} \right\}$$

4) Upper bdd

$$f_t(t) < \lim_{\substack{t \rightarrow T \\ t < T}} f_t(t) = \min\left(K, \frac{rK}{\delta}\right)$$

Call:-

1)  $f_t$  is non increasing 2) Upper bound

same as put

$$f_t(t) < \left(\frac{\lambda_1}{\lambda_1 - 1}\right) K$$

$$\lambda_1 = \frac{1}{\sigma^2} \left\{ -\left(r - \delta - \frac{\sigma^2}{2}\right) + \sqrt{\left(r - \delta - \frac{\sigma^2}{2}\right)^2 + 2\sigma^2 r} \right\}$$

3) Lower bdd

$$f_t(t) = \lim_{\substack{t \rightarrow T \\ t < T}} f_t(t) > \max\left(K, \frac{rK}{\delta}\right)$$

Put:-  $r \rightarrow 0, f_t(t) = 0 \Rightarrow$  Always hold  $\Rightarrow$  same as European  
For American call,  $\delta \rightarrow 0 \Rightarrow f_t(t) \rightarrow \infty \Rightarrow$  same as "

$$V_p^{Am}(f_t(t), t) = K - f_t(t)$$

$$\frac{\partial V_p}{\partial S}(f_t(t), t) = -1$$

$$V_c^{Am}(f_t(t), t) = f_t(t) - K$$

$$\frac{\partial V_c}{\partial S}(\cdot, \cdot) = 1$$

Finance Interpretation?

Black Scholes

$$f_{BS}(v) =$$

$$\frac{\partial v}{\partial t} +$$

$$\textcircled{1} \Rightarrow \frac{\partial v}{\partial t}$$

(From

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We know

# Black-Scholes Inequality

$$\mathcal{L}_{BS}(V) = \frac{1}{2} \sigma^2 S^2 \frac{\partial^2 V}{\partial S^2} + (r - \delta) S \frac{\partial V}{\partial S} - rV \rightarrow (1)$$

$$\frac{\partial V}{\partial t} + \left[ \frac{\sigma^2 S^2}{2} \frac{\partial^2 V}{\partial S^2} + (r - \delta) S \frac{\partial V}{\partial S} - rV \right] = 0$$

$$\frac{\partial V}{\partial t} + \mathcal{L}_{BS}(V) = 0$$

$$V = K - S \quad \frac{\partial V}{\partial t} = 0 \quad \frac{\partial V}{\partial S} = -1 \quad \frac{\partial^2 V}{\partial S^2} = 0$$

$$(1) \Rightarrow \frac{\partial V}{\partial t} + \mathcal{L}_{BS}(V) = (r - \delta) S(-1) - r(K - S) = \delta S - rK$$

$$\frac{\partial V}{\partial t} + \mathcal{L}_{BS}(V) < 0$$

(From the upper bound of Early Exercise curve)

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T.S.T  $\Delta_f$

We know that when  $t = T$ ,  $V_p(S, T) = K - S$ ,  $0 < S < K$

$$\frac{\partial V}{\partial t} + \frac{\sigma^2}{2} \frac{\partial^2 V}{\partial S^2} + (r - \delta) S \frac{\partial V}{\partial S} - rV = 0$$

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

$$\text{Claim: } \frac{\partial V}{\partial t}(S, T) \leq 0$$

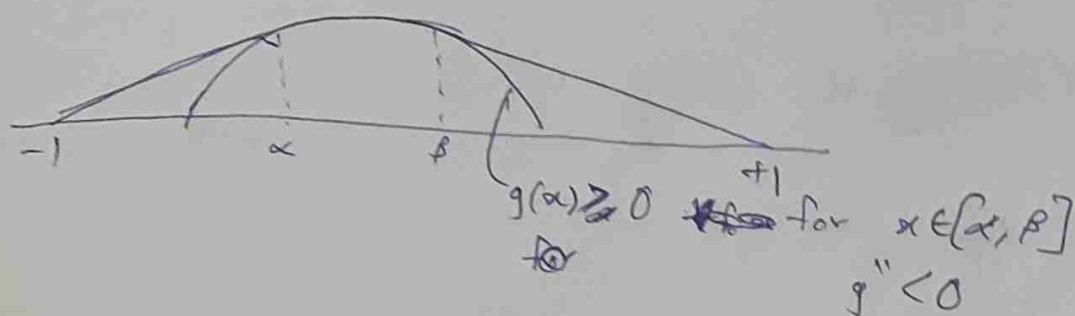
Pf:- Suppose  $\geq 0$

$$\frac{V(S, T) - V(S, T - \Delta t)}{\Delta t} \geq 0$$

$$(K - S) \Rightarrow V(S, T - \Delta t) \Rightarrow \Leftarrow$$

$$\Delta_f(T) = \lim_{\substack{t \rightarrow T \\ t < T}} \Delta_f(t)$$

## Obstacle Problem:-



Find min length of the elastic band that will keep the  $g(x)$  bowl intact

$$\begin{aligned} \text{for } -1 < x < \alpha, \quad u'' &= 0 \\ \alpha < x < \beta, \quad u &= g \\ \beta < x < 1, \quad u'' &= 0 \end{aligned}$$

$$\begin{aligned} \text{IF } u > g, \quad u'' &= 0 \\ u = g, \quad u'' &\leq 0 \quad u'' = g'' \leq 0 \\ (\text{same like American}) \end{aligned}$$

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The above can be also be rewritten as:-  
Linear Complementarity problem (LCP):-

$$\begin{aligned} u''(u - g) &= 0, \quad -u'' \geq 0, \quad u - g \geq 0 \\ u(\pm 1) &= 0 \end{aligned}$$

Constraints  
like a linear programming problem

$$(U_{i-1} - 2U_i + U_{i+1})(U_i - g_i) = 0, \quad -(U_{i-1} - 2U_i + U_{i+1}) \geq 0, \quad (U_i - g_i) \geq 0$$

$$\begin{aligned} (U - g)^T B U &= 0 \\ B U &\geq 0 \\ B U &\geq g \end{aligned}$$

$$B = \begin{pmatrix} 2 & -1 & & 0 \\ -1 & 2 & -1 & \\ & \ddots & \ddots & \ddots \\ 0 & & & \end{pmatrix}$$

In Order to solve LCP, we have to use some iterative methods to satisfy  $U_i \geq g_i \equiv$  At every iteration, make  $U_i = 0$  when  $U_i < g_i$  O.W let it be as it is while solving  $B U = 0$

BVP will be evaluated at each time level for American options i.e Black Scholes PDE will be discretized to be equivalent to LCP

Parametrized Successive OverRelaxation (PSOR)





We can reformulate LCP

$$u = u - q$$

Cryer Problem:  $y = Au - b$  ( $A$  and  $b$  are known)

Find  $x, y$  s.t.

$$\tilde{b} = b - Aq$$

$$Ax - y = \tilde{b}, x \geq 0, y \geq 0$$

$$x^T y = 0$$

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The Cryer problem (10) is equivalent to minimization for  $\min_{x \geq 0} G(x)$ ,  $G(x) = \frac{1}{2}(x^T A x) - \tilde{b}^T x$  is strictly convex.

Cryer  
AP

$$G_{xx} = A$$

$$G_{xx} = A$$

where the tridiagonal matrix  $A$

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

The eigen values of  $A$  are +ve.

$\therefore$  Hessian of  $G = G_{xx}$  is symmetric +ve definite.

$\therefore G$  is strictly convex and has a unique min on any convex set in particular for  $x \geq 0$  in  $\mathbb{R}^n$ .

From KKT,

$$\min G(x)$$

$$H_i(x) \leq 0 \quad i=1:n$$

which is equivalent to have a vector  $x_0$ .

$\exists$  Lagrangian multiplier  $y \geq 0$  s.t.

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

$$y^T H(x) = 0$$

$$(Ax - \tilde{b}) - y = 0$$

$$-y^T x = 0$$

Here we want

constraint to be  $x \geq 0$ ,  $H(x) = -x$

$$\frac{\partial H}{\partial x} = -1$$