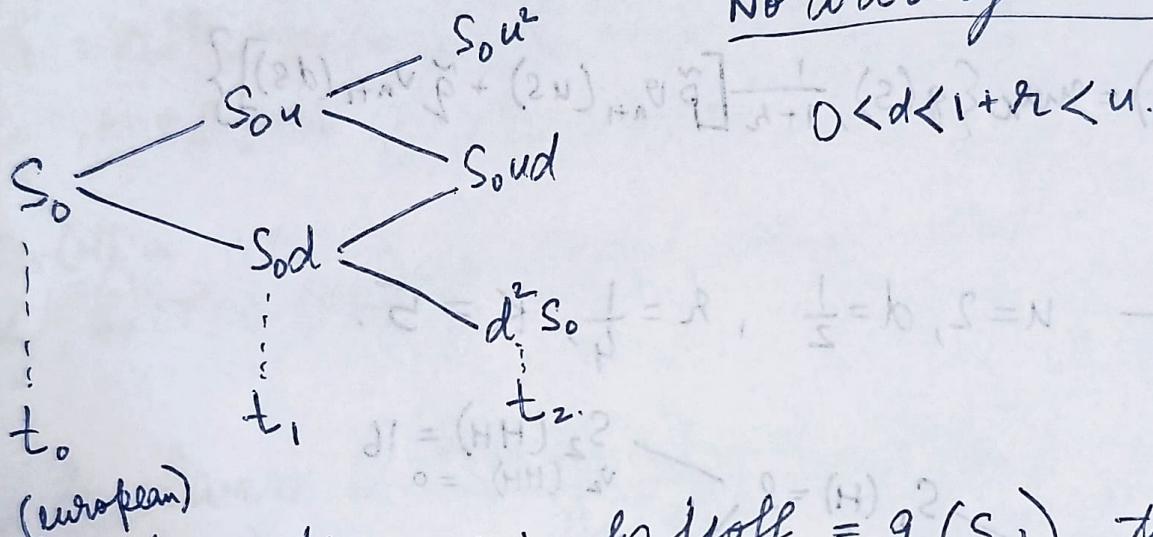


Straddles $\rightarrow f(T) = |S(T) - K|$

American Derivatives

N-period binomial model with up factor u and down factor d , interest rate r .

No arbitrage condition:-



For a derivative giving payoff $= g(S_N)$, then the price v_n at time $n \leq N$ is given by

$$\boxed{v_n(S_N) = g(S_N)}.$$

$$v_{N-1}(S_{N-1}) = \frac{\tilde{E}_{N-1}(g(S_N))}{1+r}$$

$$S_{N-1} \xrightarrow[\tilde{q}]{} dS_{N-1}$$

$$= \frac{1}{1+r} [\tilde{p} \cdot g(uS_{N-1}) + \tilde{q} \cdot g(dS_{N-1})]$$

$$v_n(S) = \frac{1}{1+r} [\tilde{p} v_{n+1}(Su) + \tilde{q} v_{n+1}(Sd)]$$

$$\text{where } \tilde{p} = \frac{1+r-d}{u-d}, \quad \tilde{q} = 1 - \tilde{p}$$

$$\text{Hedging } \Delta_n = \frac{v_{n+1}(uS_n) - v_{n+1}(dS_n)}{(u-d)S_n}$$

Suppose American derivative with payoff

$$g(S_n) \quad \forall n \leq N.$$

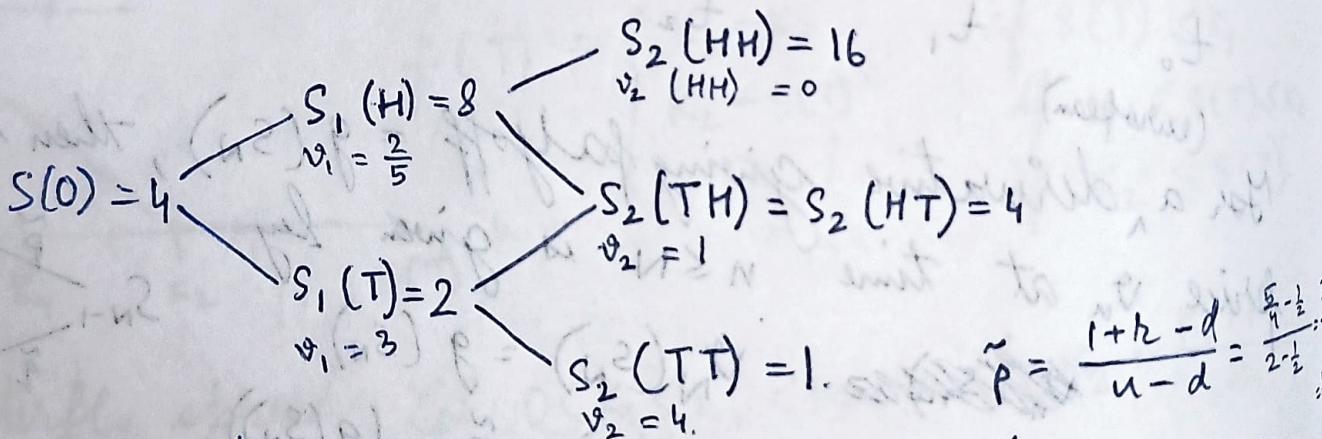
The price is given by v_n .

$$v_N(S_N) = \max \{ g(S_N), 0 \}.$$

$$v_{N-1}(S_N) = \max \{ g(S_{N-1}), \frac{1}{1+r} [\tilde{p} v_N(uS_{N-1}) + \tilde{q} v_N(dS_{N-1})] \}$$

$$v_n(S) = \max \{ g(S), \frac{1}{1+r} [\tilde{p} v_{n+1}(uS) + \tilde{q} v_{n+1}(dS)] \}.$$

Example - $u=2, d=\frac{1}{2}, r=\frac{1}{4}, K=5$.



Consider an American put with $N=2$ and $g(s) = K-s$.

$$v_2(s) = \max \{ 5-s, 0 \}$$

$$v_n(s) = \max \{ 5-s, \frac{1}{1+r} [\tilde{p} v_{n+1}(us) + \tilde{q} v_{n+1}(ds)] \}, \quad n=0,1.$$

$$v_2(16) = 0, \quad v_2(4) = 1, \quad v_2(1) = 4.$$

$$v_1(8) = \frac{1}{5/4} [v_2(0) + \frac{1}{2} \cdot 1], 0 \} = \frac{4}{5} \cdot \frac{1}{2} = \frac{2}{5}.$$

$$v_1(2) = \left\{ \frac{1}{5/4} \left[\frac{1}{2} \cdot 1 + \frac{1}{2} \cdot 4 \right], 3 \right\} = 3, \quad v_0(4) = \max \left\{ \frac{1}{5/4} \left[\frac{1}{2} \cdot \frac{2}{5} + \frac{1}{2} \cdot 3 \right], 1 \right\} = 1.36$$

$$\begin{array}{ccc}
 & v_1(8) = 0.4 & v_2(16) = 0 \\
 v_0(4) = 1.36 & & v_2(4) = 1 \\
 & v_1(2) = 3 &
 \end{array}$$

Hedging strategy -

$$\begin{aligned}
 0.4 &= \Delta_0^{(H)} S_1(H) + (1+r) (X_0 - \Delta_0^{(H)} S_0) \\
 &= 8 \Delta_0^{(H)} + \frac{5}{4} (1.36 - 4 \Delta_0^{(H)})
 \end{aligned}$$

$$\Rightarrow \Delta_0(H) = -0.43 \quad \left. \begin{array}{l} \\ \text{Always comes same.} \end{array} \right\}$$

$$\text{Similarly, } \Delta_0(T) = -0.43$$

$$\Delta_n \text{ is always } = \frac{v_{n+1}(uS_n) - v_{n+1}(dS_n)}{(u-d) S_n}$$

$$Y_n = \tilde{v}_n = \frac{1}{(1+r)^n} v_n$$

$$\begin{array}{ccc}
 Y_1(H) = 0.32 & & Y_2(HH) = 0 \\
 Y_0(4) = 1.36 & & Y_2(HT) = Y_2(TH) = 0.64 \\
 & Y_1(T) = 2.4 & Y_2(TT) = 2.56
 \end{array}$$

$$\tilde{P} = \tilde{q} = \frac{1}{2}$$

If martingale, $\langle Y_n \rangle$

$$\begin{aligned}
 \tilde{\mathbb{E}}_n[Y_{n+1}] &= Y_n \\
 \Rightarrow Y_1(T) &= \frac{1}{2} (Y_2(TH) + Y_2(TT)) = \frac{1}{2} (2.56 + 0.64) = 1.6 \neq 2.4
 \end{aligned}$$

Hence, not martingale
but super martingale.

Theorem - Consider a N -period binomial model with
 $0 < d < 1 + r < u$, $\tilde{p} = \frac{1+r-d}{u-d}$, $\tilde{q} = 1 - \tilde{p}$,
 Let payoff function $g(s)$ be given and define

$v_N(s), v_{N-1}(s), \dots, v_0(s)$ by

$$v_N(s) = \max\{g(s), 0\}.$$

$$v_n(s) = \max\{g(s_n), \frac{1}{1+r} [\tilde{p} v_{n+1}(us_n) + \tilde{q} v_{n+1}(ds_n)]\}.$$

$$\text{Define } \Delta_n = \frac{v_{n+1}(us_n) - v_{n+1}(ds_n)}{(u-d)s_n},$$

$$C_n = v_n(s_n) - \frac{1}{1+r} [\tilde{p} v_{n+1}(us_n) + \tilde{q} v_{n+1}(ds_n)].$$

$$X_{n+1} = \Delta_n s_{n+1} + (1+r)(X_n - C_n - \Delta_n s_n)$$

then,

$$x_n = v_n \quad \& \quad x_n \geq g(s_n) \quad \forall n.$$

Stopping Time

A random variable $T : \Omega_N \rightarrow \{0, 1, 2, \dots, N, \infty\}$ satisfies
 if $T(w_1, w_2, \dots, w_n, w_{n+1}, \dots, w_N) = n$, then

$$T(w_1, w_2, \dots, w_n, w'_{n+1}, \dots, w'_N) = n + w'_{n+1} - w'_N.$$

Our example:- (Stopping time when $g(s) > 0$).

$$T : \{HH, HT, TH, TT\} \rightarrow \{1, 2, \infty\}.$$

$$T(HH) = \infty$$

$$T(HT) = 2$$

$$T(TH) = T(TT) = 1.$$

stopping time

Consider $f(HH) = 0$ $f(TH) = 1$
 $f(HT) = 0$ $f(TT) = 2$... f is not a stopping time.

$$Y_{0 \wedge T} = Y_0 = 1.36$$

$$Y_{1 \wedge T} = Y_1$$

$$Y_{2 \wedge T}(HH) = Y_2(HH) = 0$$

$$Y_{2 \wedge T}(HT) = Y_2(HT) = 0.64$$

$$Y_{2 \wedge T}(TH) = Y_1(T) = 2.4$$

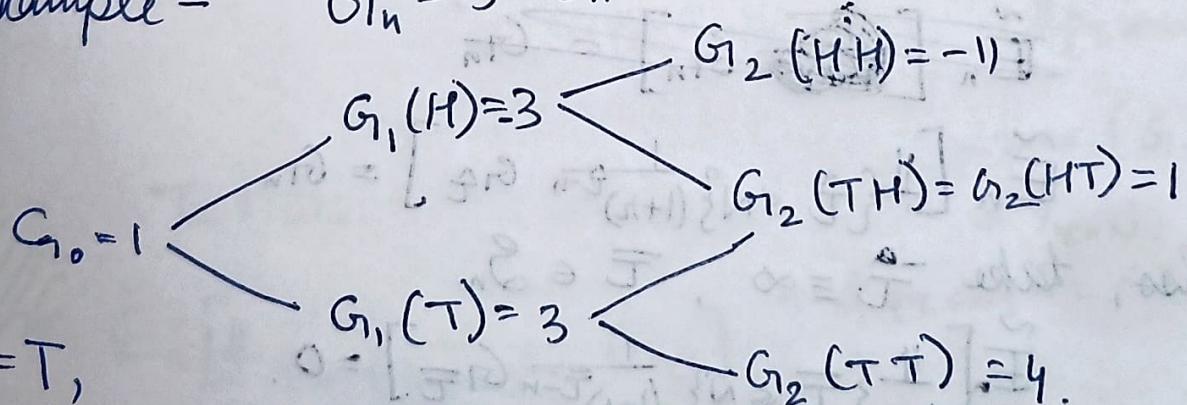
$$Y_{2 \wedge T}(TT) = Y_1(TT) = 2.4.$$

$S_n = \{T\text{-stopping time} : T : \Omega_N \rightarrow \{n, n+1, \dots, N, \infty\}\}$

Consider an American derivative with intrinsic value at n is G_n .

Define:- $V_n = \max_{T \in S_n} \tilde{E}_n \left[\mathbb{I}_{\{T \leq N\}} \frac{1}{(1+r)^{T-n}} G_{T^-} \right], n = 0, 1, \dots, N.$

Our example - $G_n = 5 - S_n$.



If $w_1 = T$,
possible T 's -

$$\tau_2(TT) = 2, \tau_2(TH) = \infty$$

$$\tau_2(TT) = 2, \tau_2(TH) = 2$$

$$\therefore \tau_2(TT) = \infty, \tau_2(TH) = \infty$$

$$\tau_2(TT) = \infty, \tau_2(TH) = 2$$

$$\tau_2(TT) = 1 = \tau_2(TH).$$

For $T(H) = 1 = T(T)$

$$\begin{aligned}V_n^T(T) &= \tilde{\mathbb{E}}_1 \left[\mathbb{I}_{\{\bar{T} \leq N\}} \frac{1}{(1+r)^{\bar{T}-1}} \right] \\&= \tilde{\mathbb{E}}_1 [G_1(T)] \\&= 3.\end{aligned}$$

General American Derivatives

Theorem - If $v_n = \max_{\bar{T} \in S_n} \tilde{\mathbb{E}}_n \left[\mathbb{I}_{\{\bar{T} \leq N\}} \frac{1}{(1+r)^{\bar{T}-n}} G_{\bar{T}} \right]$

then, ① $v_n \geq \max \{G_n, 0\}$, $\forall n$

② $\frac{1}{(1+r)^n} v_n$ is a supermartingale

③ If y_n is another process satisfying ① and ②,

then $v_n \leq y_n$.

Proof :- ① Take $\bar{T} = n$, $\bar{T} \in S_n$.

$$\tilde{\mathbb{E}}_n [\text{scratched } G_n] = G_n$$

$$\tilde{\mathbb{E}}_n \left[\mathbb{I}_{\{\bar{T} \leq N\}} \frac{1}{(1+r)^{\bar{T}-n}} G_{\bar{T}} \right] = G_n$$

Also, take $\bar{T} = \infty$, $\bar{T} \in S_n$.

$$\tilde{\mathbb{E}}_n \left[\mathbb{I}_{\{\bar{T} \leq N\}} \frac{1}{(1+r)^{\bar{T}-n}} G_{\bar{T}} \right] = 0.$$

(i.e.) $v_n \geq G_n$ and $v_n \geq 0$

$$\Rightarrow v_n \geq \max \{G_n, 0\}.$$

② Let n be given and suppose τ^* attains the maximum in the definition of V_{n+1} .

$$V_{n+1} = \tilde{E}_{n+1} \left[\mathbb{I}_{\{\tau^* \leq N\}} \frac{1}{(1+\lambda)^{\tau^*-n}} G_{\tau^*} \right].$$

$$\tau^* \in S_{n+1} \Rightarrow \tau^* \in S_n.$$

$$\therefore V_n \geq \tilde{E}_n \left[\mathbb{I}_{\{\tau^* \leq N\}} \frac{1}{(1+\lambda)^{\tau^*-n}} G_{\tau^*} \right].$$

$$= \tilde{E}_n \left[\tilde{E}_{n+1} \left[\mathbb{I}_{\{\tau^* \leq N\}} \frac{1}{(1+\lambda)^{\tau^*-n}} G_{\tau^*} \right] \right].$$

~~$\tilde{E}_n [V_{n+1}]$~~ = $\tilde{E}_n \left[\frac{1}{(1+\lambda)} V_{n+1} \right].$

$$\therefore \boxed{\frac{1}{(1+\lambda)^n} V_n \geq \tilde{E}_n \left[\frac{1}{(1+\lambda)^{n+1}} V_{n+1} \right]}$$

③ Suppose Y_n is another process satisfying ① and ②,

$$Y_k \geq \max \{G_k, 0\} \quad \forall k.$$

$$\begin{aligned} \mathbb{I}_{\{\tau \leq N\}} G_\tau &\leq \mathbb{I}_{\{\tau \leq N\}} \max \{G_\tau, 0\} \leq \mathbb{I}_{\{\tau \leq N\}} \max \{G_{\tau \wedge N}, 0\} \\ &\quad + \mathbb{I}_{\{\tau = \infty\}} \max \{G_{\tau \wedge N}, 0\} \\ &= \max \{G_{\tau \wedge N}, 0\}. \end{aligned}$$

~~$\tau \in S_n.$~~

~~$\tilde{E}_n \left[\mathbb{I}_{\{\tau \leq N\}} \frac{1}{(1+\lambda)^{\tau \wedge n}} G_\tau \right]$~~

~~$\leq \tilde{E}_n \left[\frac{1}{(1+\lambda)^{\tau \wedge n}} Y_{\tau \wedge n} \right]$~~

~~$\leq \frac{1}{(1+\lambda)^{\tau \wedge n}} Y_{\tau \wedge n} \quad (\text{From ②}).$~~

$$= \frac{1}{(1+\lambda)^n} Y_n \quad (\text{Because } \tau \in S_n \Rightarrow \tau \geq n).$$

$$\Rightarrow \tilde{E} \left[\mathbb{1}_{\{T \leq N\}} \frac{1}{(1+\lambda)^{T-n}} G_T \right] \leq y_n. \quad \text{V.T.}$$

* Lic Miss

$$[\sigma^P \frac{1}{n^{\alpha}(n+1)} \{n2^{\alpha}\}^{-1}]_{n=1}^{\infty} \leq \sqrt{N}$$

$$[\sigma^P \frac{1}{n^{\alpha}(n+1)} \{n2^{\alpha}\}^{-1}]_{n=1}^{\infty} =$$

$$[\sigma^P \frac{1}{(n+1)}]_{n=1}^{\infty} = [\sigma^P n]_{n=1}^{\infty}$$

$$[\sigma^P \frac{1}{(n+1)}]_{n=1}^{\infty} \leq \sigma^P \frac{1}{(n+1)}$$

③ Die offizielle wahl ist eine in d. wahl

$$\forall \{0, \omega\} \text{ neu} \leq \sigma^P$$

$$\{n2^{\alpha}\} \leq \sigma^P \{n2^{\alpha}\} \text{ neu}$$

④ wahl

⑤ wahl

$$[\sigma^P \frac{1}{n^{\alpha}(n+1)} \{n2^{\alpha}\}^{-1}]_{n=1}^{\infty}$$

$$[\sigma^P \frac{1}{(n+1)}]_{n=1}^{\infty} \geq 0$$

⑥ wahl: $\max^P \frac{1}{n^{\alpha}(n+1)} \geq \frac{1}{2^{\alpha}}$

($n2^{\alpha} < 2^{\alpha} T$ wahl)

Stopping Time

Definition - A random variable τ taking values in $[0, \infty]$ is called a stopping time if $\{\tau \leq t\} \in \mathcal{F}(t) \forall t$.

$$dS(t) = rS(t)dt + \sigma S(t)d\tilde{W}(t)$$

$$\text{Define } T_m = \min \{t \geq 0 : S(t) = m\}$$

If $S(t)$ never reaches level m , then we interpret $T_m = \infty$.

Ex - $X(t)$ is a martingale (sub/super) then $X(t \wedge T_m)$ is also a martingale (sub/super).

$$X(t \wedge T_m) = \begin{cases} X(t), & t \leq T_m \\ X(T_m), & t > T_m. \end{cases}$$

Perpetual put-

The underlying asset price is given by

$$dS(t) = rS(t)dt + \sigma S(t)d\tilde{W}(t)$$

where $\tilde{W}(t)$ is a BM under risk-neutral \tilde{P} .

$$S(t) = S(0) e^{\sigma \tilde{W}(t) + (r - \frac{\sigma^2}{2})t}$$

The american perpetual put: pay $\max(K - S(t), 0)$ (No T).

$K - S(t)$ if it is exercised at time t .

Def - Let \mathcal{S} be the set of all stopping time

$$V_*(x) = \max_{\tau \in \mathcal{S}} \mathbb{E}[(K - S(\tau)) e^{-r\tau}], \quad S(0) = x.$$

If $\tau = \infty$, we interpret $e^{-r\tau}$ as 0.

Lemma - Let $\tilde{W}(t)$ be a BM under $\tilde{\mathbb{P}}$. Let μ be a real number and let m be a positive number.

Set $X(t) = \mu t + \tilde{W}(t)$.

$$T_m = \min \{t \geq 0 : X(t) = m\}.$$

If $X(t)$ never reaches m , then $T_m = \infty$.

$$\tilde{\mathbb{E}}[e^{-\lambda T_m}] = e^{-m(-\mu + \sqrt{\mu^2 + 2\lambda})} > 0.$$

$$e^{-\lambda T_m} = 0 \text{ if } T_m = \infty.$$

Let a level L

If $S(0) = x < L$,

$$g_L(x) = (K - x).$$

If $S(0) = x > L$,

$$T_L = \min \{t \geq 0 : S(t) = L\}.$$

$$g_L(x) = \tilde{\mathbb{E}}[(K - S(T_L)) e^{-rT_L}].$$

$$= (K - L) \tilde{\mathbb{E}}[e^{-rT_L}].$$

$$S(t) = x e^{\sigma \tilde{W}(t) + (\lambda - \frac{1}{2}\sigma^2)t}.$$

$$S(T_L) = L.$$

$$\Leftrightarrow x e^{\sigma \tilde{W}(T_L) + (\lambda - \frac{1}{2}\sigma^2)T_L} = L.$$

$$\Leftrightarrow -\sigma \tilde{W}(T_L) + (\lambda - \frac{1}{2}\sigma^2)T_L = \ln(L/x).$$

$$\Leftrightarrow -\tilde{W}(T_L) - \frac{1}{\sigma} (\mu - \frac{1}{2}\sigma^2) T_L = \frac{1}{\sigma^2} \ln \left(\frac{m}{L} \right).$$

Here, $\mu = -\frac{1}{\sigma} (\mu - \frac{1}{2}\sigma^2)$, $\lambda = \mu$, $m = \frac{1}{\sigma} \ln \left(\frac{m}{L} \right)$.

$$\mu^2 + 2\lambda = \frac{1}{\sigma^2} \left(\mu - \frac{1}{2}\sigma^2 \right)^2 + 2\mu$$

$$= \frac{1}{\sigma^2} \left(\mu + \frac{1}{2}\sigma^2 \right)^2$$

$$\Rightarrow \mu + \sqrt{\mu^2 + 2\lambda} = \cancel{\mu + \frac{1}{2}\sigma^2} \cdot \frac{2\mu}{\sigma}$$

$$\tilde{E}[e^{-\mu T_L}] = e^{-\frac{1}{\sigma} \ln \left(\frac{m}{L} \right) \cdot \frac{2\mu}{\sigma}} = \left(\frac{m}{L} \right)^{-\frac{2\mu}{\sigma^2}}$$

$$\therefore v_L(x) = \begin{cases} (K-L) \left(\frac{x}{L} \right)^{-\frac{2\mu}{\sigma^2}}, & x > L \\ (K-x), & x \leq L. \end{cases}$$

Using $\frac{\partial v_L(x)}{\partial L} = 0$ (for maximum),

$$g(L) := (K-L) L^{2\mu/\sigma^2}$$

$$g'(L) = -L^{2\mu/\sigma^2} + \frac{2\mu}{\sigma^2} (K-L) L^{\frac{2\mu}{\sigma^2}-1} = 0.$$

$$\Rightarrow L^{2\mu/\sigma^2} = \frac{2\mu}{\sigma^2} (K-L) L^{\frac{2\mu}{\sigma^2}}$$

$$\Rightarrow L = \frac{2\mu}{\sigma^2} (K-L)$$

$$\Rightarrow L^* = \frac{\frac{2\mu}{\sigma^2} K}{1 + \frac{2\mu}{\sigma^2}} = \frac{2\mu K}{\sigma^2 + 2\mu}$$

$$\therefore v_{L^*}(x) = \begin{cases} (K-x), & \text{if } x \leq L^* \\ (K-L^*) \left(\frac{x}{L^*}\right)^{-\frac{2h}{\sigma^2}}, & x > L^*. \end{cases}$$

$$v'_{L^*}(x) = \begin{cases} -1, & x \leq L^* \\ -(K-L^*) \frac{2h}{\sigma^2} \left(\frac{x}{L^*}\right)^{-\frac{2h}{\sigma^2}}, & x > L^*. \end{cases}$$

$v'_{L^*}(x)$ is discontinuous at $x = L^*$.

$$v''_{L^*}(x) = \begin{cases} 0, & x < L^* \\ (K-L^*) \frac{2h(2h+\sigma^2)}{\sigma^4 x^2} \left(\frac{x}{L^*}\right)^{-\frac{2h}{\sigma^2}}, & x > L^*. \end{cases}$$

For $x > L^*$,

$$\begin{aligned} r v_{L^*}(x) - r x v'_{L^*}(x) - \frac{1}{2} \sigma^2 x^2 v''_{L^*}(x) \\ = \left(\frac{x}{L^*}\right)^{-\frac{2h}{\sigma^2}} (K-L^*) \left[h + \frac{2h^2}{\sigma^2} - \frac{h(2h+\sigma^2)}{\sigma^2} \right] = 0 \end{aligned}$$

If $0 \leq x < L^*$,

$$\begin{aligned} r v_{L^*}(x) - r x v'_{L^*}(x) - \frac{1}{2} \sigma^2 x^2 v''_{L^*}(x) \\ = r(K-x) + rx = rK. \end{aligned}$$

$v_{L^*}(x)$ satisfies the linear complementarity conditions:

$$v(x) \geq (K-x)^+ \quad \text{if } x \geq 0 \quad \text{--- (1)}$$

$$r v(x) - r x v'(x) - \frac{1}{2} \sigma^2 x^2 v''(x) \geq 0 \quad \text{if } x > 0 \quad \text{--- (2)}$$

and $r v(x) - r x v'(x) - \frac{1}{2} \sigma^2 x^2 v''(x) = 0$, equality holds in either (1) or (2) --- (3)

Theorem:- $\tau_{L^*} = \min \{t \geq 0 : S(t) = L^*\}$.

Then $e^{-rt} v_{L^*}(S(t))$ is a supermartingale and
 $e^{-r(t \wedge \tau_{L^*})} v_{L^*}(S(t \wedge \tau_{L^*}))$ is a martingale.

Proof:- $d(e^{-rt} v_{L^*}(S(t)))$

$$= -re^{-rt} v_{L^*}'(S(t)) dt + e^{-rt} d v_{L^*}(S(t))$$

$$= -re^{-rt} v_{L^*}'(S(t)) dt + e^{-rt} \left[v_{L^*}'(S(t)) dS(t) + \frac{1}{2} v_{L^*}''(S(t)) dS(t) dS(t) \right]$$

$$\cancel{= -re^{-rt} v_{L^*}'(S(t)) dt} +$$

$$= e^{-rt} \left[-r v_{L^*}'(S(t)) + r S(t) v_{L^*}'(S(t)) + \frac{1}{2} \sigma^2 S^2(t) v_{L^*}''(S(t)) \right] dt +$$

$$+ e^{-rt} S(t) d\tilde{W}(t)$$

$$= e^{-rt} rK \mathbb{1}_{\{S(t) < L^*\}} dt + e^{-rt} \sigma S(t) d\tilde{W}(t) v_{L^*}'(S(t)).$$

$\Rightarrow e^{-rt} v_{L^*}(S(t))$ is a supermartingale.

$$\int_0^{u \wedge \tau_{L^*}} d(e^{-rt} v_{L^*}(S(t))) = - \int_0^{u \wedge \tau_{L^*}} e^{-rt} rK \mathbb{1}_{\{S(t) < L^*\}} dt + \int_0^{u \wedge \tau_{L^*}} e^{-rt} \sigma S(t) v_{L^*}'(S(t)) d\tilde{W}(t)$$

$$\Rightarrow e^{-r(u \wedge \tau_{L^*})} v_{L^*}(S(u \wedge \tau_{L^*})) - v_{L^*}(S(0)) = \int_0^{u \wedge \tau_{L^*}} e^{-rt} \sigma S(t) v_{L^*}'(S(t)) d\tilde{W}(t).$$

$e^{-r(t \wedge \tau_{L^*})} v_{L^*}(S(t \wedge \tau_{L^*}))$ is a martingale.

Corollary -

$$v_{L^*}(x) = \max_{\tau \in S} \tilde{E} [e^{-r\tau} (K - S(\tau))], \quad (S(0) = x).$$

Proof :-

$e^{-rt} v_{L^*}(S(t))$ is a supermartingale:

$$v_{L^*}(x) = v_{L^*}(S(0)) \geq \tilde{E} [e^{-rt} v_{L^*}(S(t))].$$

Let $\tau \in S$, then $e^{-r(t \wedge \tau)} v_{L^*}(S(t \wedge \tau))$ is also supermartingale.

$$v_{L^*}(x) \geq \tilde{E} [e^{-r(t \wedge \tau)} v_{L^*}(S(t \wedge \tau))].$$

$t \rightarrow \infty$; By DCT (v_{L^*} is bdd).

$$\begin{aligned} v_{L^*}(x) &\geq \tilde{E} [e^{-r\tau} v_{L^*}(S(\tau))] \\ &\geq \tilde{E} [e^{-r\tau} (K - S(\tau))] \quad (\text{by } ①). \end{aligned}$$

$$v_{L^*}(x) \geq \max_{\tau \in S} \tilde{E} [e^{-r\tau} (K - S(\tau))].$$

Now,

$e^{-r(t \wedge \tau_{L^*})} v_{L^*}(S(t \wedge \tau_{L^*}))$ is a martingale.

$$v_{L^*}(x) = \tilde{E} [e^{-r(t \wedge \tau_{L^*})} v_{L^*}(S(t \wedge \tau_{L^*}))].$$

$t \rightarrow \infty$, by DCT,

$$v_{L^*}(x) = \tilde{E} [e^{-r\tau_{L^*}} v_{L^*}(S(\tau_{L^*}))].$$

$$= \tilde{E} [e^{-r\tau_{L^*}} (K - S(\tau_{L^*}))] \quad (\text{by } ①).$$

$$\therefore v_{L^*}(x) \leq \max_{\tau \in S} \tilde{E} [e^{-r\tau} (K - S(\tau))].$$

$$\therefore v_{L^*}(x) = \max_{\tau \in S} \tilde{E} [e^{-r\tau} (K - S(\tau))].$$

Linear Complementarity conditions -

$$g(x) \geq (K-x)^+ + x \quad \textcircled{a}$$

$$rV(x) - rxV'(x) - \frac{1}{2}\sigma^2 x^2 V''(x) \geq 0 \quad \forall x \geq 0 \quad \textcircled{b}$$

and $\forall x > 0$, equality holds in \textcircled{a} or \textcircled{b} . $\rightarrow \textcircled{c}$

$$d[e^{-rt} V_{L^*}(S(t))] = -e^{-rt} r K \mathbb{I}_{\{S(t) < L^*\}} dt \\ + e^{-rt} \sigma S(t) V'_{L^*}(S(t)) d\tilde{W}(t).$$

Corollary - Consider an agent with initial capital $X(0) = V_{L^*}(S(0))$. Suppose agent uses the portfolio

$\Delta(t) = V'_{L^*}(S(t))$ and consumed cash at the rate $c(t) = r K \mathbb{I}_{\{S(t) < L^*\}}$.

Then, $X(t) = V_{L^*}(S(t)) \quad \forall t$ and $X(t) \geq (K-x)^+$.

Proof - $dX(t) = \Delta(t) dS(t) + r(X(t) - \Delta(t) S(t)) dt - c(t) dt$.

$$d[e^{-rt} X(t)] = -r e^{-rt} X(t) dt + e^{-rt} dX(t).$$

$$= e^{-rt} [\Delta(t) dS(t) - r \Delta(t) S(t) dt - c(t) dt].$$

$$= e^{-rt} [\sigma S(t) \Delta(t) d\tilde{W}(t) - c(t) dt].$$

~~$$= \sigma e^{-rt} V'_{L^*}(S(t)) S(t) d\tilde{W}(t) - r K \mathbb{I}_{\{S(t) < L^*\}} dt.$$~~

$$= d(e^{-rt} V_{L^*}(S(t))).$$

$$\therefore e^{-rt} X(t) = e^{-rt} V_{L^*}(S(t)) \Rightarrow X(t) = V_{L^*}(S(t)).$$

Stopping set $\mathcal{I} = \{x \geq 0 : \vartheta_{L^*}(x) = (K - x)^+\}$.

Continuation set $\mathcal{C} = \{x \geq 0 : \vartheta_{L^*}(x) > (K - x)^+\}$.

Let $V(t)$ be the value process of the American perpetual put option, then

- ① $V(t) \geq (K - S(t))^+$ $\forall t \geq 0$
- ② $e^{-rt} V(t)$ is a supermartingale
- ③ \exists a stopping time T^* such that
$$V(0) = \tilde{\mathbb{E}}[e^{-rT^*} (K - S(T^*))^+]$$

American put :- $[0, T]$

Underlying asset - $dS(t) = rS(t)dt + \sigma S(t)d\tilde{W}(t)$.

Def - Let $0 \leq t \leq T$ and $x \geq 0$ be given, $S(t) = x$.

$$\mathcal{F}_u(t) = \{S(\tau) : t \leq \tau \leq u\}$$

$\mathcal{S}_{t,T}$ [set of all stopping times for $\mathcal{F}_u^{(t)}$ $t \leq u \leq T$, taking values in $[t, T]$ or ∞ .]

$$V_{t,x} = \max_{\tau \in \mathcal{S}_{t,T}} \tilde{\mathbb{E}}[e^{-r(\tau-t)} (K - S(\tau)) \mid S(t) = x].$$

$$\tau = \infty \Rightarrow e^{-r(\infty-t)} (K - S(\infty)) = 0.$$

$$[(\infty)]_v = (\infty)_X \Leftarrow (\infty)_v = (\infty)_X$$

Linear complementarity conditions-

$$v(t, x) \geq (K-x)^+ \quad \forall x \geq 0, t \in [0, T] \quad \text{--- (a)}$$

$$r v(t, x) - v_t(t, x) - r x v_n(t, x) - \frac{1}{2} \sigma^2 x^2 v_{nn}(t, x) \geq 0 \\ \forall x \geq 0, t \in [0, T] \quad \text{--- (b)}$$

$\forall t \in [0, T], x \geq 0$ equality holds either in (a) or (b).

$$\mathcal{I} = \{(t, x) : v(t, x) = (K-x)^+\}$$

$$\mathcal{C} = \{(t, x) : v(t, x) > (K-x)^+\}$$

$$T^* = \min \{ u \in [t, T] : v(u; S(u)) = (K-S(u))^+ \}.$$

Theorem - $e^{-rt} v(t, S(t))$ is a supermartingale. $t \leq u \leq T$.
 and the stopped process $e^{-r(u \wedge T^*)} v(u \wedge T^*, S(u \wedge T^*))$ is a martingale.

Corollary - Consider an agent with initial capital $X(0) = v(0, S(0))$. Suppose this agent uses the portfolio process:
 $\Delta(u) = v_n(u, S(u))$ and consume at rate

$$C(u) = r K \mathbb{1}_{\{S(u) \leq L(T-u)\}}. \quad \text{Then } X(u) = v(u, S(u)) \quad \forall u \geq 0 \\ \text{and } X(u) \geq (K-S(u))^+.$$

Continuous time Portfolio Theory

S_t^i = price of 1 unit of i^{th} asset at time t

h_t^i = no. of units of i^{th} asset at t .

h_t^n = portfolio process $(h_t^1, h_t^2, \dots, h_t^N)$.

V_t^n = value of portfolio $(h_t^1, h_t^2, \dots, h_t^N)$ at t .

$$V_t^n = h_t^n S_t$$

$$dV_t^n = h_t^n dS_t = \sum_{i=1}^N h_t^i dS_t^i$$

Defn - For given portfolio h , corresponding relative portfolio or portfolio weights w is given by

$$w_t^i = \frac{h_t^i S_t^i}{V_t^n}, \quad i=1, 2, \dots, N$$

$$\sum_{i=1}^N w_t^i = 1$$

Lemma - A portfolio is self-financing if

$$\begin{aligned} dV_t^n &= \sum_{i=1}^N h_t^i dS_t^i \\ &= V_t^n \sum \frac{w_t^i dS_t^i}{S_t^i} \end{aligned}$$

Defⁿ:- Consider a financial market with vector price process & contingent claim (with date of maturity T also called T-claim) is any RV $X \in \mathbb{F}_t^S$.

A contingent claim is called simple if $X = \phi(S(T))$. The function ϕ is called the contract fn.

e.g., European call $\rightarrow \max\{x - K, 0\} = \phi(x)$.

Notation:- $\Pi(t, X)$ for price process of the contingent claim X .

$$\Pi(T, X) = X = \phi(S(T)) \text{ (for simple claim).}$$

Defⁿ:- An arbitrage possibility on a financial market is an self-financing portfolio $h \geq$ $V^h(0) = 0$, $\hat{P}(V^h(T) \geq 0) = 1$, $P(V^h(T) > 0) > 0$.

Property 1:- Suppose \exists a portfolio h ^{self-financing}

$$dV^h(t) = K(t)V^h(t)dt$$

where K is adapted. Then $K(t) = r(t) + t$.

$$dB(t) = rB(t)dt$$

$$dS(t) = S(t)\mu(t, S(t))dt + S(t)\sigma(t, S(t))dW(t).$$

$$\text{Take } X = \phi(S(T))$$

We assume that the price process is given by

$$\Pi(t, X) = F(t, S(t)) = \Pi(t).$$

$$d\pi(t) = dF(t, S(t)) = F_t(t, S(t))dt + F_x(t, S(t))dS(t)$$

$$+ \frac{1}{2} F_{xx}(t, S(t)) ds(t) dS(t).$$

$$= (F_t + \mu S F_x + \frac{1}{2} S^2 \sigma^2 F_{xx}) dt + S \sigma F_x dW(t).$$

$$d\pi(t) = \mu_\pi(t)\pi(t)dt + \sigma_\pi(t)\pi(t)dW(t),$$

where $\mu_\pi(t) = \frac{F_t + \mu S F_x + \frac{1}{2} S^2 \sigma^2 F_{xx}}{\cancel{F}}$

and $\sigma_\pi(t) = \frac{S \sigma F_x}{\cancel{F}}$

Consider a portfolio (u_S, u_π)

$$dV(t) = V(t) \left[\frac{u_S dS(t)}{S(t)} + \frac{u_\pi d\pi(t)}{\pi(t)} \right]$$

$$= V(t) [u_S \mu + u_\pi \mu_\pi] dt + V(t) [u_S \sigma + u_\pi \sigma_\pi] dW(t)$$

$$u_S + u_\pi = 1.$$

$$\begin{cases} u_S \sigma + u_\pi \sigma_\pi = 0 \\ u_S + u_\pi = 1 \end{cases} \Rightarrow u_S = \frac{\sigma_\pi}{\sigma_\pi - \sigma}, \quad u_\pi = \frac{-\sigma}{\sigma_\pi - \sigma}$$

$$dV(t) = V(t) [u_S \mu + u_\pi \mu_\pi] dt.$$

$$u_S \mu + u_\pi \mu_\pi = r.$$

$$u_S = \frac{\sigma_\pi}{\sigma_\pi - \sigma} = \frac{\sigma S F_x / F}{\sigma S F_x / F - \sigma} = \frac{S F_x}{S F_x - F}$$

$$u_\pi = \frac{-\sigma}{F_\pi - \sigma} = \frac{-F}{S F_x - F}$$

$$\frac{SF_n}{SF_n - F} \mu + \frac{-F}{SF_n - F} \left[\frac{F_t + \mu SF_n + \frac{1}{2}\sigma^2 S^2 F_n x}{F} \right] = r$$

$$\Rightarrow \boxed{F_t + r S F_n + \frac{1}{2}\sigma^2 S^2 F_n x = r F}$$

$$\pi(T, \phi) = \phi(S(T))$$

$$F(T, x) = \phi(x)$$

Meta-theorem

Let M denote the number of underlying traded assets in the model (excluding the risk-free asset) and let R denote the number of random sources. Then,

- ① The model is arbitrage free iff $M \leq R$
- ② The model is complete iff $M \geq R$
- ③ The model is complete & arbitrage free iff $M = R$.