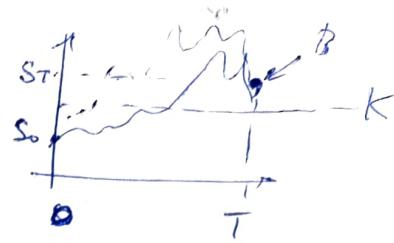


Lecture 1

- 1) - call
- put
- Barrier
- asian

1) (Call) Barrier option

$$\text{Payoff} = f(S_t), \quad (S_t)_{t \in [0, T]}$$



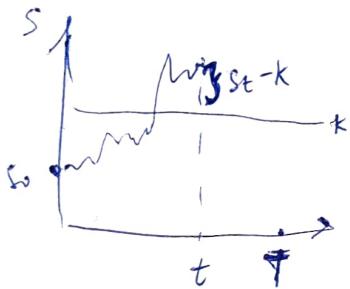
American option

$$(A_T - K)^+$$

↑ average of prices of S

American option (can be exercised

at any time before mat.)



Problem.

1) Price

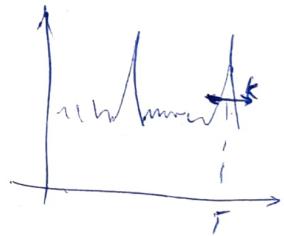
2) Hedging.

(fin.) 3) Exercise? Optimal Stopping
+ related Optimisation problem

* price of electricity

has jumps

⇒ by taking
average it
can be normalized



2) Principles of financial models

a) value of time $X_T = X_t (1 + r(T-t))$

If we assume r fixed $\rightarrow S_t e^{rt} = S_0 e^{r(T-t)}$
risk neutral probability $\rightarrow Q = \frac{S_t e^{rt} - S}{S - S_0}$ \rightarrow this is true iff $S < e^{rt} S_0 < S^*$
THE IEP $\rightarrow X_T = e^{r(T-t)} X_t$ $\Rightarrow X_t = e^{-r(T-t)} X_T$

Article:

very process-driven
mean reverting electricity
price model;

$X_T = e^{-r(T-t)} X_t$
discounted value of X_T

b) risk-neutral pricing $S_0 = (p S^* + (1-p) S^-) e^{-rt}$ \leftarrow risk-neutral price
 \rightarrow Asset S_0 is always invested and produces money for the bank

$S_T = S^- + p e^{rT} S^* - (1-p) S^-$

$$\text{Option } C := e^{-rt} \cdot (qC^+ + (1-q)C^-)$$

$$\sigma = \begin{cases} (S+K)^+ \\ (S-K)^+ \end{cases}$$

$\Rightarrow q$ [risk-neutral probability] \Downarrow
 \Rightarrow [risk-neutral price]

$$= f(S_0, r, T, K, S^+, S^-)$$

observable params
↓

choosing by
calibration

to make a model
realistic

• P - real world prob.

When working with TS, we are estimating

- q - is pricing measure (\Rightarrow risk neutral measure)

• No-arbitrage rule

Implication:
knowing

$X_T = Y_T \Rightarrow X_t = Y_t$, $\forall t \leq T$ (otherwise we can buy cheaper one $\Rightarrow \dots$)

(Ex)

$$X_t = S_t + P_t$$

$$Y_t = C_t + K \cdot e^{-r(T-t)}$$

bank

$$X_T = \max(S_T, K)$$

$$Y_T = \max \{ S_T - K, 0 \} + K = X_T$$

N.A. Principle

$$\Rightarrow S_t + P_t = C_t + K \cdot e^{-r(T-t)}$$

obs. Put-Call parity obs.

should satisfy, else
we can arbitrage and
make money for free

$$V = \alpha S + \beta B$$

← bank account

$V_t = G_t$ → find replicating strategy

(Ex)

$$\begin{cases} \alpha S^+ + \beta e^{rt} = C^+ \\ \alpha S^- + \beta e^{rt} = C^- e^{-r \cdot \text{decrease}} \end{cases}$$

↑ price increase
↑ strategy

$$\Rightarrow \begin{cases} \bar{\alpha} = \frac{C^+ - C^-}{S^+ - S^-} \\ \bar{\beta} = \dots \end{cases}$$

equality
holds

	\circ	T
Bank	\circ	e^{rt}
Asset	S_0	$S = \begin{cases} S^+ \\ S^- \end{cases}$
Price	$C_0 = ?$	$G = \begin{cases} (S^+ - K)^+ \\ (S^- - K)^+ \end{cases}$

$$\boxed{V_T = C_T} \Rightarrow \text{no-arbitrage price } C_0 := \bar{\alpha} S_0 + \bar{\beta} \cdot 1 = V_0$$

if price $\neq C_0$ $\Rightarrow C_0 = f(S_0, T, \kappa, \rho, S^+, S^-)$
 market has an arbitrage opportunity

N.A. principle gives hedging solution for bank

* C_0 price for an option equal to money V_0 that needs to be invested

To find V_0 we have lin. system by bank strategy and cond. $V_T = C_T$

* * risk-neutral price - the expectation of the payoff like Cauchy pr.

Incomplete Markets

	0	T
Bank	1	e^{rt}
Asset	S_0	$S_T = \begin{cases} S^1 \\ S^2 \\ S^3 \end{cases}$
option.	C_0	$q_1 \\ q_2 \\ 1-q_1-q_2$

$$R-N \quad S_0 = e^{-rT} (q_1 S^1 + q_2 S^2 + (1-q_1-q_2) S^3)$$

1 eq 2 unk.
 \Rightarrow inf 2-n

measures
 (probabilities)

\Rightarrow incomplete model

(because mart. measure isn't unique)

$$\begin{cases} dS^1 + \beta B = C^1 \\ dS^2 + \beta B = C^2 \\ dS^3 + \beta B = C^3 \end{cases}$$

in general $\not\exists (\alpha, \beta)$

not solvable

C_1, C_2, C_3 and only 2 params (d.o.f.) $\Rightarrow \not\exists N.A. \text{ price}$
 three conditions

\Rightarrow incomplete model

To complete market another derivative is ~~needed~~ needed

(Ω, \mathcal{F}, P)
 ↑ ↗ prob. measure
 σ -algebra
 (information)

1) Measurability (deterministic dependence)

$$X: \Omega \rightarrow \mathbb{R}^d$$

$$\forall A \in \mathcal{B} \quad (X \in A) \in \mathcal{F}$$

$\mathcal{G}(x) \leftarrow \sigma$ -algebra about x

Problem: measurability $x \in m \mathcal{Y} \Leftrightarrow \mathcal{G}(x) \subseteq \mathcal{Y}$

Assume $y \in m \mathcal{G}(x) \Leftrightarrow y = f(x)$
 $f \in m \mathcal{B}$

X, Y independent w.r.t. P might become dependent

if we change the p. measure

Lecture 2

$$\textcircled{1} \quad P(A|B) = \frac{P(A \cap B)}{P(B)}$$

$$P(A|\delta(B))(w) = \begin{cases} P(A|B), & \text{if } w \in B \\ P(A|B^c), & \text{if } w \in B^c \end{cases}$$

$$\textcircled{2} \quad g = \delta(y)$$

$$Y(\Omega) = \{y_1, \dots, y_N\} \text{ or } \{y_k\}_{k \in \mathbb{N}}$$

and all elements
are unions
of partition
of space

$$E[X|Y]^w = E[X|\delta(Y)](w) = E[X|g = y(w)]$$

event

$$\textcircled{3} \quad W = \boxed{E[X|g] = z} \quad \begin{matrix} \text{a.s.} \\ \text{s.t.} \\ \text{general} \\ \text{b-algebra} \end{matrix}$$

$$1) z \in mg$$

$$2) E[zw] = E[Xw] \quad \forall w \in mg$$

not possible
for Borel
algebra

$$L^2(\Omega, \mathcal{F}, P) : E[X^2] < +\infty \quad (\text{like Euclidean space for inf. dimensions})$$

$\begin{matrix} \xrightarrow{\quad} \\ L^2(\Omega, \mathcal{G}, P) \\ \text{subspace} \end{matrix}$

$\begin{matrix} \xleftarrow{\quad} \\ L^2(\Omega, \mathcal{F}, P) \\ \text{space} \end{matrix}$

Interpretation of (2):
orthogonality

$$\langle z - x, w \rangle_{L^2} = E[(z-x) \cdot w] = 0$$

Prop.

$$1) E[X] = E[Z]$$

$$2) X \in mg \Rightarrow X = E[X|g] \rightarrow \text{best case est.}$$

$$3) X, g \text{ independent w.r.t. } P \Rightarrow E[X|g] = E[X] \Rightarrow \text{the worst expectation}$$

$$4) E[XY|g] = y E[X|g], \quad y \in mg \quad \text{we can set for } X$$

knowledge
of const.
smaller by

$$5) g \subseteq H$$

$$\begin{aligned} E[E[X|g]|H] &= \\ &= E[E[X|H]|g] = \\ &= E[X|g] \end{aligned}$$

(Ex) $X(\Omega) = \{X_n\}_{n \in \mathbb{N}}$
 $Y(\Omega) = \{Y_n\}_{n \in \mathbb{N}}$

if $P(X = x_n | Y)$ is constant $\forall n$
 $\Leftrightarrow X, Y$ ind. w.r.t. P

Martingales: $(\Omega, \mathcal{F}, P, (\mathcal{F}_n)_{n \in \mathbb{N}})$

$(X_n)_{n \in \mathbb{N}}$ \uparrow filtration
 $\hookrightarrow \mathcal{F}_n \leq \mathcal{F}_{n+1} \dots \leq \mathcal{F}_N$

$\mathcal{F}_n^X = \sigma(X_1, \dots, X_n)$ ← natural filtration of X

$X = (X_n)_{n \in \mathbb{N}}$

if $X_n \in \mathcal{F}_n \quad \forall n$ then X is adapted to $(\mathcal{F}_n)_{n \geq 1}$ adapted

MG: $X = (X_n)_{n \geq 1}$ on $(\Omega, \mathcal{F}, P, (\mathcal{F}_n))$

if $X_n = E[X_{n+1} | \mathcal{F}_n] \quad \forall n$

Remarks:

- 1) X is MG $\Rightarrow X$ adapted
- 2) $X_n = E[X_{n+1} | \mathcal{F}_n] =$
 $= E[E[X_{n+2} | \mathcal{F}_{n+1}] | \mathcal{F}_n]$ (not on trajectories!)
 $= E[X_{n+2} | \mathcal{F}_n]$

$\Rightarrow X_n = E[X_{n+k} | \mathcal{F}_n], \forall k \geq 1$

3) $E[X_n] = E[X_{n+k}] = E[X_0]$

(Ex) $(Y_k)_{k \geq 1}$ $(\Omega, \mathcal{F}, P, (\mathcal{F}_n^X))$

$Y_k \stackrel{i.i.d.}{=} \text{Ber}(p) \Rightarrow Y_k = \begin{cases} 1 & , p \\ 0 & , 1-p \end{cases}$

$$X_n = \sum Y_k$$

is X a MG?
 $E[X_{n+1} | \mathcal{F}_n^X] = E[X_n + Y_{n+1} | \mathcal{F}_n^X] =$
 $= X_n + E[Y_{n+1} | \mathcal{F}_n^X]$ independent $= X_n + E[Y_{n+1}] = X_n + \frac{2p-1}{2}$

$$p = 1/2$$

If $p > 1/2$ $X_n \leq E[X_{n+1} | \mathcal{F}_n^X]$ X is a sub-mg.
 $(X_n \geq E[X_{n+1} | \mathcal{F}_n^X] \Rightarrow \text{super-mg})$

Poet's Decomposition (\Rightarrow Ho formula)

X adapted process drift (gives direction)

$$X_n = M_n + A_n$$

\uparrow
predictable
 $A_0 = 0$

$t_n \in \mathbb{N}$ $\forall n$

$A_n \in m\mathcal{F}_{n-1}$ $t_n \geq 1$

$$X_{n+1} - X_n = \underbrace{M_{n+1} - M_n}_{\star} + A_{n+1} - A_n$$

$$E[X_{n+1} | \mathcal{F}_n] - X_n = 0 + A_{n+1} - A_n \Rightarrow \begin{cases} A_0 = 0 \\ A_{n+1} = A_n + E[X_{n+1} | \mathcal{F}_n] - X_n \end{cases}$$

is uniq. defined

(Ex) $M_n = X_n - n(2p-1)$
 $A_n = n(2p-1)$

Market Model



$$t_n = k \cdot \frac{T}{N}, \quad k = \overline{0, N}$$

Bank

$$\begin{cases} B_0 = 1 \\ B_{n+1} = B_n (1+r) \\ \Rightarrow B_n = (1+r)^n \end{cases}$$

\uparrow
price at t_n

s^1, \dots, s^α - assets

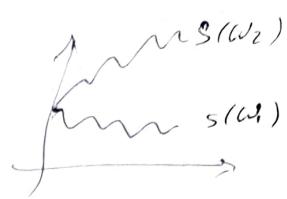
$$\begin{cases} S_{n+1}^i = S_n^i (1+\mu_{n+1}^i) \\ S_0^i > 0 \end{cases}$$

\uparrow
rate of return

$$\mu = (\mu_n)_{n \geq 0} \text{ s.p. on } (-\infty, \infty, \mathcal{F}^M)$$

strategy : $(\alpha_n, \beta_n)_{n=1, \dots, h}$

α is d -dimensional



$$V_n = \underbrace{\alpha_n \cdot S_n}_{\sum_{i=1}^d \alpha_n^i S_n^i} + \beta_n \cdot B_n$$

$w, w_2 \in \mathbb{R}$

$$\sum_{i=1}^d \alpha_n^i S_n^i$$

1) (α, β) are predictable *

$$(\alpha_n, \beta_n) \in m \gamma_{h-1}, \forall n \geq 1$$

2) self-financing

$$V_n = \alpha_{n+1} \cdot S_n + \beta_{n+1} \cdot B_n, \quad \forall n \quad (\text{can't draw money from outside})$$

$\hat{V} = h(\alpha, \beta)$ predictable + self-financing }

↳ set of admissible strategies

discounted

$$\text{lem: } \tilde{V}_n = \alpha_{n+1} \tilde{S}_n + \beta_{n+1} \cdot 1$$

$$\tilde{V}_{n+1} = \alpha_{n+1} \tilde{S}_{n+1} + \beta_{n+1} \cdot 1$$

$$\tilde{V}_{n+1} - \tilde{V}_n = \alpha_{n+1} (\tilde{S}_{n+1} - \tilde{S}_n)$$

$\frac{\partial V}{\partial S}$ $\frac{\partial \tilde{V}}{\partial \tilde{S}}$

self-fin \Rightarrow value ch. $\frac{\partial V}{\partial S}$ dep. only on $\frac{\partial S}{\partial \tilde{S}}$ price ch.

• knowing V_0 and pred. α
 $\Rightarrow (\alpha, \beta)$

Model is correct?

(DEF) Arbitrage opportunity : $(\alpha, \beta) \in A$ s.t. 1) $V_0^{(\alpha, \beta)} = 0$

$$2) V_N^{(\alpha, \beta)} \geq 0 \text{ a.s.}$$

$$3) P(V_N^{(\alpha, \beta)} > 0) > 0$$

(REM) : $Q \sim P$ equivalent measures (e.g. same events has eq. measure)

Model : arbitrage-free \Rightarrow good model

if and do not exist

(DEF) Martingale Measure - prob. measure $Q \sim P$ s.t.

\tilde{S} is a Q -mg

discounted
 \Rightarrow time change effect
is removed

$$\Rightarrow \tilde{S}_n = E^Q [\tilde{S}_{n+1} | \mathcal{F}_n]$$

risk-neutral measure

1^o Fund. Theor. of asset pricing

* A model is arbitrage-free iff a mg measure exists

Prop. $(\alpha, \beta) \in A$ and Q - mg. measure

then $\tilde{V}^{(\alpha, \beta)}$ is a Q -mg

$$\begin{aligned} E^Q [\tilde{Y}_{n+1} - \tilde{V}_n | \mathcal{F}_n] &= E^Q [\alpha_{n+1} (\tilde{S}_{n+1} - \tilde{S}_n) | \mathcal{F}_n] \stackrel{*}{=} \\ &= \alpha_{n+1} \underbrace{E^Q [\tilde{S}_{n+1} - \tilde{S}_n | \mathcal{F}_n]}_{=0} = 0 \end{aligned}$$

\Rightarrow a martingale condition

N.A. principle: $(\alpha, \beta), (\alpha', \beta') \in A$ and a mg measure exists

Then if $V_N^{(\alpha, \beta)} = V_N^{(\alpha', \beta')}$

Then $\tilde{V}_N^{(\alpha, \beta)} = V_N^{(\alpha', \beta')}$, $\forall n \leq N$

$$\tilde{V}_n^{(\alpha, \beta)} = E^\Phi [\tilde{V}_N^{(\alpha, \beta)} | \mathcal{F}_n] =$$

$$= E^\Phi [\tilde{V}_N^{(\alpha', \beta')} | \mathcal{F}_n] = \tilde{V}_n^{(\alpha', \beta')}$$

Proof to 1°:

1° \Leftrightarrow Assume $\exists Q$ s.t. if (α, β) is an arbitrage
 $0 = \tilde{V}_0^{(\alpha, \beta)} = E^\Phi [\tilde{V}_N^{(\alpha, \beta)}]$ ≥ 0
 . nonneg. ≥ 0
 . str. pos w. str. pos p

\Rightarrow impossible \Rightarrow arb. doesn't exist

\Rightarrow

European option: X r.v.

(ex) call-option $X = (S_N - k)^+$ payoff
 $= \psi(S_N)$

1) risk-neutral pricing.

$$E^\Phi \left[\frac{X}{B_N} \right]$$

not unique
 (because mg measure
 is not unique)

2) N.A. pricing

\Rightarrow interval of prices
 (R.N. price \in int)

Lecture 3

x - price

Assume: $\rightarrow Q$ market measure

1) $E^Q \left[\frac{x}{B_n} \right]$ initial price

2) $A_x^+ = \{(\alpha, \beta) \in A \mid V_n^{(\alpha, \beta)} \geq x\}$

$A_x^- = \dots \exists$

$A_x^+ : A_x^-$

$V_0^{(\alpha, \beta)} \in "price" < V_0^{(\alpha, \beta)}$

$V_{(\alpha, \beta)} \in A_x^- \quad \nexists (\alpha, \beta) \in A_x^+$

Lemma: $\sup_{(\alpha, \beta) \in A_x^-} V_n^{(\alpha, \beta)} \leq E^Q \left[\frac{x}{B_n} \right] \leq \inf_{(\alpha, \beta) \in A_x^+} V_0^{(\alpha, \beta)}$ + Pmg meas

If $(\alpha, \beta) \in A_x^+$: $E^Q \left[\frac{x}{B_n} \right] \leq E^Q \left[V_n^{(\alpha, \beta)} \right] = V_0^{(\alpha, \beta)}$

Reducing: find $(\alpha, \beta) \in A$ s.t. $V_n^{(\alpha, \beta)} = x$ if payoff is replicable

$\Rightarrow (\alpha, \beta) \in A_x^+ \cap A_x^-$

strategy is super and sub repl.

\Rightarrow price is equal to risk-neutral price

Complete model: $\nexists x$ is replicable
(enor. option)

↓
II Fund. Theory of asset pricing: a model is complete iff
the mg measure is unique

Binomial model : $d=1$

$$S_n = (1+r)^n$$

(Ω, \mathcal{F}, P)

$$\begin{cases} S_0 > 0 \\ S_{n+1} = S_n(1 + \mu_{n+1}) \end{cases}$$

$1 + \mu_n \sim \text{Ber}(u, d)$, $0 < d < u$

$\text{if } \mu_1, \dots, \mu_n - \text{i.i.d} \Rightarrow P \text{ is uniq. def.} :$

$$P(S_n = u^k d^{n-k} S_0) = \binom{n}{k} p^k (1-p)^{n-k}$$

Prop: $\exists! Q \text{ iff } d < 1+r < u$

Proof: $E^Q [S_{n+1} | \mathcal{I}_n] = S_n ?$

$$E^Q [S_{n+1} | \mathcal{I}_n] = S_n(1+r)$$

$$E^Q [S_n(1 + \mu_{n+1}) | \mathcal{I}_n] = S_n(1+r)$$

$$E^Q [1 + \mu_{n+1} | \mathcal{I}_n] = 1+r$$

$$u \otimes (1 + \mu_{n+1} = u(\mathcal{I}_n) + d(1 - \phi(\cdot))) = 1+r$$

$$Q(1 + \mu_{n+1} = u(\mathcal{I}_n)) = \frac{1+r-d}{u-\alpha} = q$$

↑
ind.
↓ $\mathcal{I}_0, \mathcal{I}_1, \dots$

$$\text{if } \mathbb{I} \otimes \Rightarrow \mu_1, \dots, \mu_n - \text{i.i.d}$$



$$Q(1 + \mu_{n+1} = u)$$

All options T, \bar{k}

$$C_0 = E^Q \left[\frac{(S_T - \bar{k})^+}{(1+r)^n} \right] = \frac{1}{(1+r)^n} \sum_{k=0}^n (u^n d^{n-k} S_0 - \bar{k})^+ \cdot Q(S_T = u^n d^{n-k} S_0)$$

$$\binom{n}{k} q^k (1-q)^{n-k}$$

Calibration a, d, r, N, σ ^{fixed}



$$\delta = \frac{T}{N} \text{ const}$$

$$1+r = e^{rs}$$

r - annual rate

$(\bar{S}_n)_{n=1, \dots, M}$ - TS - realization of s. process $\Rightarrow \bar{\mu}, \bar{\sigma}^2$

$$S_T = S_0 e^{\mu T} \Rightarrow \bar{\mu} := \frac{1}{T} \lg \frac{S_T}{S_0} = \frac{1}{T} \lg \prod_{n=0}^{N-1} (1 + \mu_{n+1})$$

$$= \frac{1}{T} \sum_{n=0}^{N-1} \lg (1 + \mu_{n+1})$$

$$\mathbb{E}[\mu] = \frac{1}{T} \underbrace{N(p \lg u + (1-p) \lg d)}_{=\frac{1}{6}} \leftarrow \text{theor. exp.}$$

$\bar{\mu} \rightarrow \text{negligible}$

$$\begin{cases} \delta \bar{\mu} = p \lg u + (1-p) \lg d \\ \delta \bar{\sigma}^2 = \dots \end{cases} \Rightarrow \mu = e^{\bar{\mu} - \frac{\delta \sqrt{\delta}}{2} + O(\delta)}, \quad S \rightarrow 0^+$$

$$\text{var}^p \left(\sum_{n=0}^{N-1} \lg (1 + \mu_{n+1}) \right) \stackrel{\text{ind.}}{=} N \cdot \text{var} (\lg (1 + \mu_1)) = N p / (1-p) \left(\lg \frac{u}{d} \right)^2$$

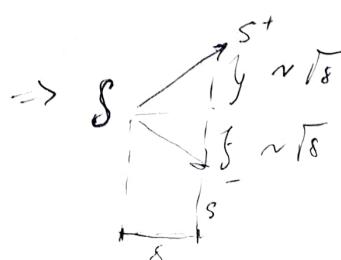
$$\Rightarrow \mu = e^{\delta \sqrt{\delta}} = \frac{1}{d}$$

δ - volatility

by Taylor exp. $\mu \approx 1 + \delta \sqrt{\delta} + O(\delta)$

$$\frac{S_{t+\delta} - S_t}{\delta} \approx \frac{\sqrt{\delta}}{\delta} \rightarrow \infty \Rightarrow \text{not differentiable}$$

process



$$d \approx 1 - \delta \sqrt{\delta} + O(\delta)$$

$$\text{Rem 1} \cdot S_N^{(\max)} = u_N^N \quad S_0 = e^{-\delta N \sqrt{\delta}} \xrightarrow[N \rightarrow \infty]{} \infty$$

$$S_N^{(\min)} \xrightarrow{} 0$$



$$\text{Rem 2} \cdot d_n < 1 + r_n < u_n$$

$$e^{-\delta \sqrt{\delta}} < e^{R\delta} < e^{\delta \sqrt{\delta}}$$

$$-\delta < R\sqrt{\delta} < \sigma$$

$\downarrow N \rightarrow \infty \Rightarrow$ model is art. -free

$$\text{Rem 3} \cdot S_T^{(n)} = S_0 \prod_{k=1}^n (1 + \mu_k^{(n)}) = S_0 e^{x^{(n)}} \xrightarrow[\text{i.i.d.}]{\text{CLT}} \underbrace{\sum_{k=1}^n \log(1 + \mu_k^{(n)})}_{\text{conv. weakly to norm. value}} \sim N\left(\frac{\delta^2}{2} T, \sigma^2 T\right)$$

B.S. formula

$$\mathbb{E} \int [S_0 e^x - K]^+ f(x) dx$$

R

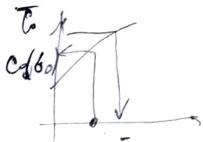
$$\begin{aligned} C_0 &= \mathbb{E}[(S_0 e^x - K)^+] \\ &\stackrel{n}{\longrightarrow} C_0 \\ \mathbb{E}[y(x^{(n)})] &\stackrel{n}{\longrightarrow} \mathbb{E}[y(x)] \\ &\quad K \in C_0 \end{aligned}$$

B.S. formula:

$$\int f(S_0, T, R, K, \delta) \quad f(S_0) > 0$$

observable?

2) implicit calibration



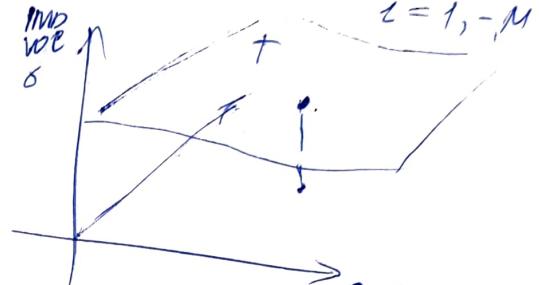
$\bar{\sigma}$ implied volatility (BS-implied volatility)

$$\bar{\sigma} = \bar{\sigma}(S_0, T, K, R)$$

Lecture 4

BS formula $e^{-RT} \cdot E[(S_0 e^X - K)^+]$

$$= \underbrace{C_{BS}(R, T, \frac{S_0}{K}, \sigma)}_{\text{P(8)}} \underbrace{\int_R^\infty}_{\text{moneyness}} \underbrace{\int_R^\infty \pi(y, t) \phi(y) dy}_{\text{gaussian payoff}}$$

Fix R : $\delta_i \leftarrow \bar{C}(T_i, k_i)$ 

American options: $(X_n)_{n=\overline{0, N}}$; adapted; $X_n \geq 0$, $X_n \in \mathcal{M}_n$

e.g.: call: $X_n = (S_n - K)^+$

$X_n \in \mathcal{M}_n$

Q8 1) Pricing $\pi_n \geq X_n$

2) Hedging: (for European): find $(\alpha, \beta) \in A$, s.t. $V_n^{(\alpha, \beta)} = X_n$ (payoff)

- Find $(\alpha, \beta) \in A$, s.t. $V_n^{(\alpha, \beta)} = X_n$, π_n

but $\pi_{(\alpha, \beta)} \neq \pi$

(even in complete market models). bcs

~~V~~ $V^{(\alpha, \beta)}$ is a Q -mgf X_n - is generic process

+ Find $(\alpha, \beta) \in A$ s.t. $V_n^{(\alpha, \beta)} \geq X_n$, π_n

early exercise strategy

3) Optimal early exercise:

stepping time: $\tau: \mathbb{N} \rightarrow \{1, \dots, N\}$ z.v.

s.t. $(\tau = n) \in \mathcal{F}_n, \pi_n$

{w.e.l. $\tau(w) = n\}$

$(X_\tau)(w) = X_{\tau(w)}(w)$, $w \in \Omega$ w.v. - a payoff if exercise $\tau(w)$ at τ .

$$\tilde{X}_t = \frac{X_t}{B_t}$$

$$\sup_{t \in \mathbb{T}} E^{\Phi} [\tilde{X}_t] = E^{\Phi} [\tilde{X}_{\tau_0}]$$

To is optimal if
given

$$\left\{ \begin{array}{l} A_x^+ = \{ (\alpha, \beta) \in A \mid \\ V_h^{(\alpha, \beta)} \geq X_h \text{ for } \forall \end{array} \right.$$

$$A_x^- = \{ (\alpha, \beta) \in A \mid$$

$$\exists t \in \mathbb{T} \text{ s.t. } V_t^{(\alpha, \beta)} \leq X_t \}$$

$$\sup_{(\alpha, \beta) \in A_x^+} V_0^{(\alpha, \beta)} \leq \text{price } h_0 \leq \inf_{(\alpha, \beta) \in A_x^-} V_0^{(\alpha, \beta)}$$

If $h_0 > V_0^{(\alpha, \beta)}$ with $(\alpha, \beta) \in A_x^+$
sell buy

\Rightarrow arbitrage, because of A_x^+

If $\frac{h_0}{\text{buy}} < \frac{V_0^{(\alpha, \beta)}}{\text{sell}}$ for some $(\alpha, \beta) \in A_x^+$

\Rightarrow arbitrage

Lemma: if Φ exists then

$$\sup_{t \in \mathbb{T}} E^{\Phi} [\tilde{X}_t] = h_0$$

any price \Rightarrow N.A price

► based on optional sampling theorem

a) if $(\alpha, \beta) \in A_x^-$

$$E^{\Phi} [\tilde{V}_t^{(\alpha, \beta)}] \leq E^{\Phi} [\tilde{X}_t] \leq \underbrace{\sup_{\tau_1 \in \mathbb{T}} E^{\Phi} [\tilde{X}_{\tau_1}]}_{\text{is independent of } (\alpha, \beta)}$$

by OST $\tilde{V}_0^{(\alpha, \beta)} = E^{\Phi} [\tilde{V}_t^{(\alpha, \beta)}]$ because at this time t is s.t. $\tilde{V}_t^{(\alpha, \beta)} \leq h_0$

\tilde{H}_n - price process

Def: $\tilde{H} = M + A \leftarrow$ s.t. $\tilde{H}_0 = 0$, $A \downarrow$
 s.t. $\tilde{H} = M_0$

M_n = European option

By completeness

$$\Rightarrow \tilde{V}_n^{(\alpha, \beta)} = M_n + x_n$$

$$\exists (\alpha, \beta) \in A$$

goal:

$$(\alpha, \beta) \in A_x^+ \cap A_x^-$$

$$\text{s.t. } \tilde{V}_N^{(\alpha, \beta)} = M_N \Rightarrow \tilde{V}_N^{(\alpha, \beta)} = M_N \geq \tilde{H}_N \geq \tilde{x}_N$$

\Rightarrow strategy is superreplicating (A_x^+)

$$A_n = \sum (E^\Phi [\tilde{H}_n | \tilde{\mathcal{F}}_{n-1}] - \tilde{H}_{n-1})$$

$$\tilde{H}_n = \max \{ \tilde{x}_n, E^\Phi (\tilde{H}_{n+1} | \tilde{\mathcal{F}}_n) \}$$

$$T_0 = \min \{ n | \tilde{x}_n \geq \tilde{E}_L \} \quad T_1 = \min \{ n | \tilde{x}_n > \tilde{E}_n \}$$

$$A_n = 0, \quad n \leq T_0$$

$$\tilde{V}_{T_0}^{(\alpha, \beta)} = M_{T_0} = \tilde{H}_{T_0} = \tilde{x}_{T_0} \Rightarrow \dots A_x^-$$

5) If $(\alpha, \beta) \in A_x^+$ take $\tau \in \mathcal{T}$

$$\tilde{X}_\tau \leq \tilde{V}_c^{(\alpha, \beta)}$$

$$E^\Phi[\tilde{X}_\tau] \leq E^\Phi[\tilde{V}_c^{(\alpha, \beta)}]$$

$$= \tilde{V}_0^{(\alpha, \beta)}$$

by OST

upper bound doesn't dep on τ

$$\Rightarrow \sup_{\tau \in \mathcal{T}} E^\Phi[\tilde{X}_\tau] \leq \inf_{(\alpha, \beta) \in A_x^+} \tilde{V}_0^{(\alpha, \beta)}$$

N.A. price

Theory: If the market is complete ($\exists (\alpha, \beta)$)

then $\exists (\alpha, \beta) \in A_x^+ \cap A_x^-$

i) $V_n^{(\alpha, \beta)} \geq X_n \quad \forall n$

ii) $\exists t_0 \in \mathcal{T}$ s.t. $V_t^{(\alpha, \beta)} \geq X_{t_0}$

iii) $V_0^{(\alpha, \beta)} = \sup_{\tau \in \mathcal{T}} E^\Phi[\tilde{X}_\tau]$
 $\stackrel{\text{OST}}{=} E^\Phi[\tilde{V}_0^{(\alpha, \beta)}] = E^\Phi[\tilde{X}_{t_0}]$ $\Rightarrow t_0$ is optimal

(R.K.) (S_n) -adapted process

$$\tau' = \max \{n \mid S_n \notin [a, b]\}$$

$$\tau = \min \{n \geq 0 \mid S_n \notin [a, b]\}$$

τ is a stopping time

$$(\tau = n) = (S_0 \in [a, b]) \wedge \dots \wedge (S_n \in [a, b])$$

* Doob's decomposition

* Snell envelope (= smallest super-mg above...)

$$f_{t,n} = \max_{\tilde{x}_n} \left\{ \tilde{x}_n + E^\Phi \left[\frac{\tilde{E}_n}{\tilde{H}_{n+1}/f_{n+1}} \right] \right\} \quad f_{t,n} \geq \tilde{x}_n, \quad f_{t,n} \geq E^\Phi[\tilde{H}_{n+1}/f_{n+1}]$$

Stochastic Processes

1) Probability space (Ω, \mathcal{F}, P)

Gen. Theory

\mathbb{R}

$[0, 1]$

- Ω - sample space $(1, 0) \# \Omega = 2^n$ $[0, 1]$
- \mathcal{F} - σ -algebra \mathcal{F} - powerset $\# \mathcal{F} = 2^{\#\Omega} = 2^{2^n}$ $[\alpha, \beta], (\alpha, \beta], (\alpha, \beta), \{\beta\}$ \Rightarrow Borel σ -algebra
 - 1) $\Omega \in \mathcal{F}$
 - 2) $A \in \mathcal{F} \Rightarrow \Omega \setminus A \in \mathcal{F}$
 - 3) $A_1, \dots, A_n \in \mathcal{F}$
 - $\Rightarrow \bigcup_{i=1}^{\infty} A_i \in \mathcal{F}$
- P - prob. measure $P\{1\} = p, P\{0\} = 1-p, P\{[\alpha, \beta]\} = \beta - \alpha$
 - 1) $P\{\Omega\} = 1$
 - 2) $A_1, \dots, A_n \in \mathcal{F}$
 - 3) $P\{\bigcup A_i\} = \sum P\{A_i\}$

$$P : \mathcal{F} \rightarrow [0, 1]$$

Random variable: measurable function $\xi : \Omega \rightarrow \mathbb{R}$ s.t.
 $\forall B \in \mathcal{B}(\mathbb{R}) : \xi^{-1}(B) \in \mathcal{F}$

T - time

$X : T \times \Omega \rightarrow \mathbb{R}$ - random function

if $\forall t \in T : X(t, \cdot)$ - random variable on (Ω, \mathcal{F}, P)

$$T = N(\mathbb{R}) \leftarrow \begin{array}{l} T = \mathbb{R}_+ (\mathbb{R}) \\ \text{r.p. w.r.t.} \end{array}$$

\Rightarrow random (stoch.) process (c.t.)

$$T = \mathbb{R}_+^n$$

\Rightarrow random (stoch.) field

Trajectory (path) $T \rightarrow \mathbb{R}$
 fix w

Finite-dim. distr. $(X_{t_1}, X_{t_2}, \dots, X_{t_n}), t_1, \dots, t_n \in \mathbb{R}$

$$\textcircled{1.8} \quad X_t = \xi \cdot t \quad \xi = \begin{pmatrix} 1 & t_1 \\ 2 & t_2 \\ 1 & t_3 \end{pmatrix} \quad \begin{array}{c} \text{min} \left(\frac{x_1}{t_1}, \frac{x_2}{t_2} \right) < 1 \\ \text{---} \\ \text{---} \end{array}$$

$$P\{X_{t_1} \leq x_1, X_{t_2} \leq x_2\} = \begin{cases} 0 & \min \left(\frac{x_1}{t_1}, \frac{x_2}{t_2} \right) < 1 \\ \frac{1}{2} & -1 < \epsilon \leq 1 \\ 1 & -1 < \epsilon < 2 \end{cases}$$

any s.p. at any fixed time point is a random variable

(e.s.) S.p. $X_t = \sin t \xi_2 + \cos t \xi_2$, $t \geq 0$, $\xi_1, \xi_2 \sim i.i.d. N(0, 1)$

assume $A, B \in \mathbb{R}$ A for ξ_1 , B for ξ_2

$$X_t = A \sin t + B \cos t$$

if $A^2 + B^2 = 0 \Rightarrow$ the trajectory of X_t is horizontal line $X_t = 0$

$$\begin{aligned} \text{if } A^2 + B^2 \neq 0 \Rightarrow X_t &= A \sin t + B \cos t = \sqrt{A^2 + B^2} \left(\sin t \frac{A}{\sqrt{A^2 + B^2}} + \cos t \cdot \frac{B}{\sqrt{A^2 + B^2}} \right) \\ &= \sqrt{A^2 + B^2} \left(\sin t \cos(\arccos \frac{A}{\sqrt{A^2 + B^2}}) + \cos t \sin(\arccos \frac{A}{\sqrt{A^2 + B^2}}) \right) \\ &= \sqrt{A^2 + B^2} \left(\sin(t + \arccos \frac{A}{\sqrt{A^2 + B^2}}) \right) \end{aligned}$$

2) Renewal process

$$S_0 = 0$$

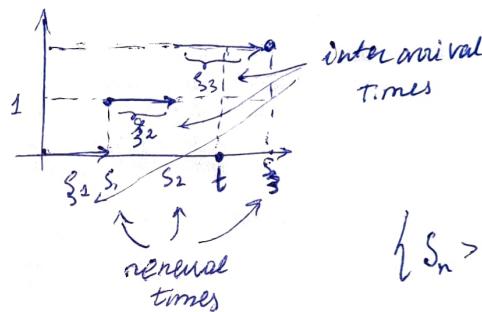
$$S_n = S_{n-1} + \xi_n, \quad \xi_1, \dots, \xi_n \text{ i.i.d. } > 0 \text{ a.s.}$$

$$\star = S_{n-1} + \xi_{n-1} + \xi_n = \sum_{k=1}^n \xi_k \quad \Leftrightarrow \mathbb{P}\{\xi_k > 0\} = 1 \quad \Leftrightarrow F(0) = 0$$

Counting Process

$$N_t = \arg \max_k \{S_k \leq t\}$$

$$F \rightarrow \mathbb{E} N_t$$



$$\{S_n > t\} = \{N_t < n\}$$

$$\text{e.g. } \{S_3 > t\} = \{N_t < 3\}$$

before time t
renewal time less than 2
3 after t events occurred