

MC Basics

Note 3

P1

[Ref]: [Gla03] chap 1

ex1 (comp)

Estimate π by the Algo below:

[Algo(N)]

① Draw N random points uniformly in rectangle $(-1, 1) \times (-1, 1)$:

$$\{Z_i = (X_i, Y_i) : X_i \sim U(-1, 1), Y_i \sim U(-1, 1) \text{ iid}, 1 \leq i \leq N\}$$

② Identify the number N_i of points inside the unit circle

$$\textcircled{3} \quad \hat{\pi}_N = 4 \cdot \frac{N_i}{N}$$



Rk π is deterministic,

$\hat{\pi}$ is random to approx. π .

Def A random (estimator) $\hat{\alpha}$ to a deterministic quantity α is called MC.

$$\textcircled{1} \text{ Bias} = E[\hat{\alpha}] - \alpha$$

$$\textcircled{2} \text{MSE}(\hat{\alpha}) = E[(\hat{\alpha} - \alpha)^2]$$

prop

$$\text{MSE}(\hat{\alpha}) = |\text{Bias}|^2 + \text{Var}(\hat{\alpha})$$

pf - - -

Rk The smaller MSE, the better MC

Def MC is called unbiased, if $\text{Bias} = 0$

ex2

$\textcircled{1}$ Find 1st, 2nd moments of $\hat{\pi}$

$\textcircled{2}$ Find Bias & MSE of $\hat{\pi}$.

Def Given a series of estimator of α , say $\{\alpha_n, n \in \mathbb{N}\}$, we say (α_n) is consistent if $p\text{-}\lim_{n \rightarrow \infty} \alpha_n = \alpha$.

(Rk) $p\text{-}\lim$ and $L\text{-}\lim \longrightarrow$ Next page

prop If $\text{MSE}(\hat{\alpha}_n) \rightarrow 0$, then $\hat{\alpha}_n$ is consistent.

ex Prove $(\hat{\pi}_n)$ is consistent estimator to π .

Def " $P\text{-}\lim_n \alpha_n = \alpha$ ", or " $\alpha_n \xrightarrow[n \rightarrow \infty]{} \alpha$ in prob." if p3

$$\lim_{n \rightarrow \infty} P(|\alpha_n - \alpha| > \varepsilon) = 0 \quad \forall \varepsilon > 0$$

Def " $L_2\text{-}\lim_n \alpha_n = \alpha$ " or " $\alpha_n \xrightarrow[n \rightarrow \infty]{} \alpha$ in L_2 " if

$$\lim_{n \rightarrow \infty} E|\alpha_n - \alpha|^2 = 0$$

ex Justify that

" $P\text{-}\lim_n \alpha_n = \alpha$ " implies " $L_2\text{-}\lim_n \alpha_n = \alpha$ "

Ans No. b/c counter-example.

Let $P \sim U([0, 1])$

$\alpha_n: [0, 1] \rightarrow \mathbb{R}$ s.t.

$$\alpha_n(\omega) = \begin{cases} n & \text{if } 0 < \omega < \frac{1}{n^2} \\ 0 & \text{otherwise} \end{cases}$$

$\alpha: [0, 1] \rightarrow \mathbb{R}$ s.t.

$$\alpha(\omega) \equiv 0$$

□

ex Justify:

" $L_2\text{-}\lim_n \alpha_n = \alpha$ " implies " $P\text{-}\lim_n \alpha_n = \alpha$ "

Ans Yes.

$\forall \varepsilon > 0$.

$$P\left(|\alpha_n - \alpha| > \varepsilon\right) = P\left(\frac{|\alpha_n - \alpha|}{\varepsilon} > 1\right)$$

$$\leq E \frac{|\alpha_n - \alpha|^2}{\varepsilon^2} \xrightarrow{n \rightarrow \infty} 0.$$

Rk
In the above we use chebyshev inequality:

$$P(|X| > 1) \leq E|X|^2.$$

§ Geometric Asian Option for BSM, P1

Bond r

stk $BSM(\sigma) \sim S_t$

i.e. $S_t = S_0 \exp\left\{\left(r - \frac{1}{2}\sigma^2\right)t + \sigma W_t\right\}$ w.r.t. \mathbb{Q}
 $\approx S_0 \exp\left\{\left(r - \frac{1}{2}\sigma^2\right)t + \sigma\sqrt{t} Z\right\}$,
 where $Z \sim N(0,1)$

~~GAC pay~~

Payoff for geometric asian option

① $GAC(T, K, n)$

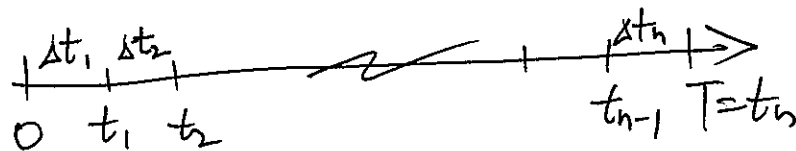
$$\pi_T^C = (A_T - K)^+$$

② $GAP(T, K, n)$

$$\pi_T^P = (A_T - K)^-$$

where

$$A_T = (S_{t_1} S_{t_2} \dots S_{t_n})^{\frac{1}{n}}$$



$$\Delta t_n = \Delta t_{n-1} = \dots = \Delta t_1 = \frac{T}{n}$$

Q.

what's e^{-rT}
 $\pi_0^C = E^{\mathbb{Q}}[\pi_T^C]$, $\pi_0^P = E^{\mathbb{Q}}[\pi_T^P]$?

$$\text{set } \mu = r - \frac{1}{2}\sigma^2, \quad \Delta t = \frac{T}{n}$$

R

$$S(t_1) = S_0 \exp(\mu \Delta t + \sigma \sqrt{\Delta t} Z_1)$$

$$S(t_2) = S(t_1) \exp(\mu \Delta t + \sigma \sqrt{\Delta t} Z_2)$$

\vdots

$$S(t_n) = S(t_{n-1}) \exp(\mu \Delta t + \sigma \sqrt{\Delta t} Z_n)$$

$$\text{where } \{Z_1, \dots, Z_n\} \text{ iid } N(0, 1)$$

$$S(t_i) = S_0 \cdot \exp\left(i\mu \Delta t + \sigma \sqrt{\Delta t} \sum_{j=1}^i Z_j\right)$$

$$\prod_{i=1}^n S(t_i) = S_0^n \cdot \exp\left(\mu \Delta t \sum_{i=1}^n i + \sigma \sqrt{\Delta t} \sum_{i=1}^n \sum_{j=1}^i Z_j\right)$$

$$= S_0^n \exp\left(\frac{(n+1)n}{2} \mu \Delta t + \sqrt{\sum_{j=1}^n (n+1-j)} \sigma \sqrt{\Delta t} Z_j\right)$$

$$\text{Note } \sum_{j=1}^n (n+1-j) Z_j \sim \text{normal with}$$

$$E\left[\sum_{j=1}^n (n+1-j) Z_j\right] = 0$$

$$= \frac{n(2n+1)(n+1)}{6}$$

$$\text{Var}\left(\sum_{j=1}^n (n+1-j) Z_j\right) = \sum_{j=1}^n (n+1-j)^2 = \sum_{j=1}^n j^2$$

$$\text{i.e. for some } \hat{Z} \sim N(0, 1)$$

$$\hat{A}(T) = S_0^n \exp\left(\frac{(n+1)n}{2} \mu \Delta t + \sqrt{\sum_{j=1}^n j^2} \sigma \sqrt{\Delta t} \hat{Z}\right)$$

$$A(T) = S_0 \exp\left(\underbrace{\frac{(n+1)}{2} \mu \Delta t}_{\hat{\mu}} + \underbrace{\sqrt{\frac{\sum_{j=1}^n j^2}{n}}}_{\hat{\sigma}} \sigma \sqrt{\Delta t} \hat{Z}\right)$$

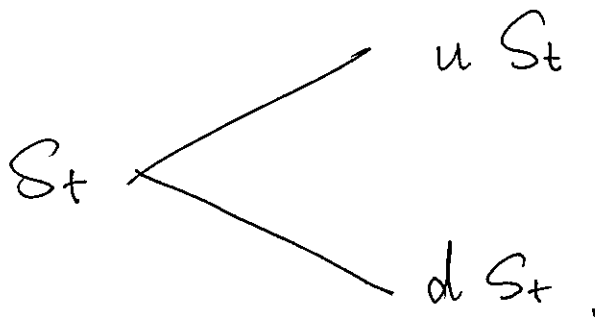
$$A(T) = S_0 \exp \left(\underbrace{\frac{(n+1)\mu}{2}}_{\hat{r} - \frac{1}{2}\hat{\sigma}^2} \Delta t + \underbrace{\frac{\sigma \sqrt{\sum_{j=1}^n j^2}}{n}}_{\hat{\sigma}} \sqrt{\Delta t} \hat{Z} \right)$$

$$\begin{aligned} \Pi_0^c &= e^{-rT} \mathbb{E}^Q \left[\left(S_0 \exp \left\{ \left(\hat{r} - \frac{1}{2} \hat{\sigma}^2 \right) \Delta t + \hat{\sigma} \sqrt{\Delta t} \hat{Z} \right\} - K \right)^+ \right] \\ &= e^{-rT} e^{\hat{r} \Delta t} \mathbb{E}^Q \left[e^{-\hat{r} \Delta t} \left(\quad \quad \quad \right)^+ \right] \\ &= e^{\hat{r} \Delta t - rT} \text{BSM-Call} (S_0, \hat{r}, \hat{\sigma}, \Delta t, K) \end{aligned}$$

CRR-model

P.

$$\left\{ \begin{array}{l} M^{\text{CRR}} = \{(\Omega, \mathcal{F}, \mathbb{F}, \mathbb{P}), T, (S, B)\} \\ t \in \{0, \Delta t, \dots, M\Delta t = T\} \triangleq \overline{T} \\ B_t = e^{-r(T-t)} \\ S_{t+\Delta t} = S_t \cdot m_t = \begin{cases} u S_t \\ d S_t \end{cases} \\ \text{where } u = e^{\sigma\sqrt{\Delta t}}, \quad d = e^{-\sigma\sqrt{\Delta t}}, \quad \sigma > 0 \end{array} \right.$$



Def $\{X_t : t \in \overline{T}\}$ is m.t.g.l. if

$$\mathbb{E}[X_{t+1} | X_t] = X_t$$

ex Find a prob \mathbb{Q} s.t. $\{e^{-rt} S_t, t \in \overline{T}\}$ is m.t.g.l.

Sols

$$q = \mathbb{Q}(m_t = u) = \frac{\cancel{e^{rt}} e^{r\Delta t} - d}{u - d}$$

$$1 - q = \mathbb{Q}(m_t = d)$$

Q why $C_n^M \rightarrow C_0$ when $M \rightarrow \infty$?

Def \mathbb{Q} is risk-neutral prob (EMM). if $\{e^{-rt} S_t, t \in \overline{T}\}$ is \mathbb{Q} -mtgl.

ex For CRR find

$$\text{Var}^{\mathbb{Q}}(\ln M_t) \text{ and } E^{\mathbb{Q}}(\ln M_t)$$

Soln $V^{\mathbb{Q}} = \sigma^2 \Delta t.$

$$E^{\mathbb{Q}} = \dots$$

ex For what σ , is $\{\ln M_t, t \in \overline{T}\}$ a \mathbb{Q} -mtgl?

Soln only if $r = \sigma$

Computation

① G is known at maturity i.e.

$$G \in \mathcal{F}_T.$$

~~② $C_t = e^{-r(T-t)}$~~

③ Compute Backwardly, i.e.

$$C_t = e^{-r \Delta t} E_t^{\mathbb{Q}}[C_{t+1}].$$

~~Ex Bin Tree Eu Option Pricing~~Ex Bin Tree Eu Option PricingGiven① Bond : $r = 0.05$

② Stk :

Bin Tree ($S_0 = 50$, $N = 2$, $T = 0.5$, $P_u = 1.2$, $P_d = 0.8$)

③ Payoff :

a) Call (T, K)

$$\overline{P}_T = (S_T - K)^+$$

b) Put (T, K)

$$\overline{P}_T = (S_T - K)^-$$

Goal

Find

~~\overline{P}_T~~ ~~P_0~~

Class
~~class~~

" Binomial European Option "

Setup-parameters:

$M = N + 1$: # of terminal nodes

$u = \textcircled{u} p_u$: up ~~grow~~ factor

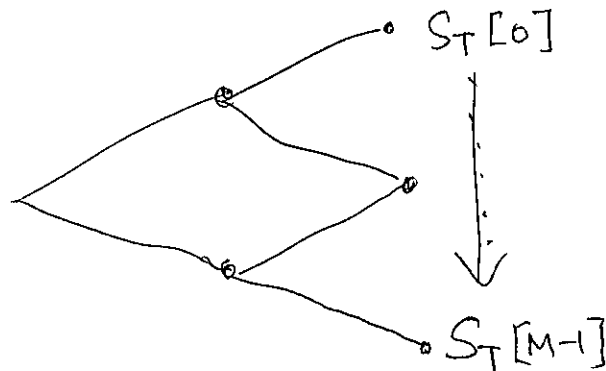
$d = \textcircled{d} p_d$: down factor

$$\left. \begin{aligned} q_u &= \frac{e^{(r-\delta)\Delta t} - d}{u - d} \\ q_d &= 1 - q_u \end{aligned} \right\} : \text{EMM. } (\delta = 0)$$

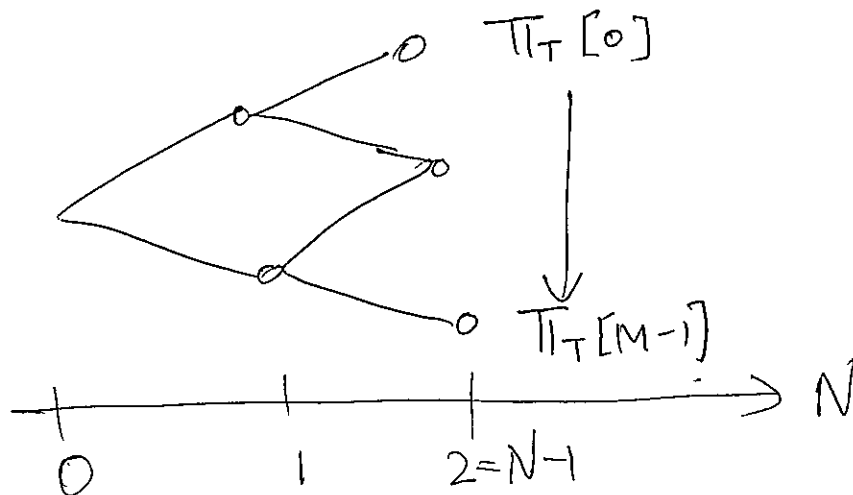
$\Delta t = \frac{T}{N}$: Time length of one period.
(inherited)

$df = e^{-(r-\delta)\Delta t}$: Discount factor for one period.

Initialize-stock-tree : Compute $\{S_T[i], i=0, 1, \dots, M-1\}$



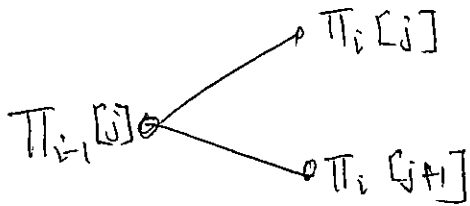
Initialize-payoff-tree : Compute $\pi_T[i], i=0 \dots M-1$



Traverse - tree : (Backward)

$$\pi_N[:] \rightarrow \pi_{N-1}[] \rightarrow \dots \rightarrow \pi_0[]$$

By EMM averaging repeatedly, i.e.



$$\pi_{i-1}[j] = (q_u \cdot \pi_i[j] + q_d \pi_i[j+1]) \cdot d_j$$

{ Option Pricing (BinTree + CRR)

P1

★ ^{ex1} BinTree ($S_0 = 50$, $N = 2$, $T = 0.5$, $u = 1.2$, $d = 0.8$, $r = 0.05$)

EuCall (T, K)

EuPut (T, K)

AmCall (T, K)

AmPut (T, K)

$T = 0.5$, $K = 50$.

Goal ① BinTree + EuCall.

② BinTree + AmPut

★

CRR ($S_0 = 50$, $N = 2$, $T = 0.5$, $\sigma = 0.3$, $r = 0.05$)

= BinTree ($S_0 = 50$, $N = 2$, $T = 0.5$, $u = e^{\sigma\sqrt{\Delta t}}$, $d = 1/u$, $r = 0.05$)

where $\Delta t = \frac{T}{N}$

Q. In ex1. find equivalent σ for CRR.

Q. EuCall (T, K) + CRR ($S_0 = 50$, $N = 2$, $T = 0.5$, $\sigma = 0.3$, $r = 0.05$)

$\Downarrow N \rightarrow \infty$

EuCall (T, K) + BSM (S_0, T, σ, r).

Q. For Bin(S_0, N, T, u, d, r)

R

① Find Q .

② Find

$$E^Q[S_{t+\Delta t} | S_t] = e^{r\Delta t} S_t \triangleq \bar{r} S_t$$

and

$$\text{Var}^Q[S_{t+\Delta t} | S_t] = \sqrt{(u - \bar{r})(\bar{r} - d)} S_t^2$$

Q Do the same for CRR(S_0, N, T, σ, r)

① Find Q

② Find $E^Q[S_{t+\Delta t} | S_t]$, $\text{Var}^Q(S_{t+\Delta t} | S_t)$

③ $\text{Var}^Q\left(\ln \frac{S_T}{S_0}\right) = \sigma^2 T$.