

## About homework

1) Robust the VB code in Excel

Sheet 2, 3 1) Optimize the BinomialTree code

Sheet 2, 3 2) Add Trinomial Tree <sup>both</sup> Strike pricing and Option pricing

Sheet 3 3) Add Black-Scholes Option pricing

Sheet 1 4) Add Greeks calculator

Sheet 3 5) Add Residual tree of Binomial Option prices - Black-Scholes option prices

Sheet 3 6) Add Mean Square Error of Binomial Option prices - Black-Scholes option price

2) Results (Analysis)

1)  $\begin{cases} \text{Price Call} \ll \text{Price Put} & \text{When Strike} > S_0 \\ \text{Price Call} > \text{Price Put} & \text{When Strike} = S_0 \\ \text{Price Call} > \text{Price Put} & \text{When Strike} < S_0 \end{cases}$

2) As n increased three prices convergent

3)  $C - P = S_0 - e^{-rT} K$  put-call parity

4) Price Binomial < Price Trinomial < Price BSM  
There are underestimations in Binomial and Trinomial Model

Analysis 1) Why the velocity of  $V_C$  converging to 0 so quicker than  $V_P$  when Strike  $> S_0$ ?

Considering BSM model, when  $K > S_0$ ,  $d_2 < d_1 < 0$

$$as \quad V_P = K e^{-r(T-t)} N(-d_2) - S N(-d_1)$$

$$V_C = -K e^{-r(T-t)} N(d_2) + S N(d_1)$$

$$\text{Whence} \quad N(d_{1,2}) \approx 0 \quad N(-d_{1,2}) \approx 1$$

$$\text{So} \quad V_C \approx 0$$

It's a very rough proof, just point out qualitative analysis.

2) Rigor Proof, see below (6 pages)

3) Easy to prove when it is Binomial or Trinomial condition  
Skip that conditions, let's see BSM model =

$$V_C - V_P = -K e^{-r(T-t)} [N(d_2) + N(d_1)] + S \pi (N(d_1) + N(-d_1))$$

As  $N(\pi) = 1 - N(-\pi)$ , substitute to equation before,

$$V_C - V_P = -K e^{-r(T-t)} + S \pi = S - e^{-rT} K$$

4) ~~BSM is the~~ Just the fact Binomial tree and Trinomial tree are approximate models of BSM. See below.  
(discrete)

Lemma 8

Preparation

Symmetric Random Walk

$$M_k = \sum_{j=1}^k X_j$$

$$\text{Var}[M_0 - M_k] = k$$

Martingale

$$\mathbb{E}[M_t | \mathcal{F}_k] = M_k$$

$$[M, M]_k = k$$

Scaled Symmetric Random Walk  $W^{(n)}(t) = \frac{1}{\sqrt{n}} M_{nt}$



$$\text{Var}[W^{(n)}(t) - W^{(n)}(s)] = t-s$$

$$\mathbb{E}[W^{(n)}(t) | \mathcal{F}_s] = W^{(n)}(s)$$

$$[W^{(n)}, W^{(n)}](t) = t$$

$n \rightarrow \infty$ , the distribution of  $W^{(n)} \sim N(0, t)$

Binomial Tree

$$1) u_n = 1 + \frac{\tau}{\sqrt{n}} \quad d_n = 1 - \frac{\tau}{\sqrt{n}} \quad r=0 \quad \text{risk-neutral}$$

$$\Rightarrow S(t) = S(0) \exp \left\{ \tau W - \frac{1}{2} \tau^2 t \right\}$$

2)

Geometric Brownian Motion  $u_n = e^{\tau/\sqrt{n}}$   $d_n = e^{-\tau/\sqrt{n}}$   $r_n = \frac{\tau}{\sqrt{n}}$  risk-neutral probabilities

$$\tilde{p}_n = \frac{\left(\frac{\tau}{\sqrt{n}} + 1\right) - e^{-\tau/\sqrt{n}}}{e^{\tau/\sqrt{n}} - e^{-\tau/\sqrt{n}}} \quad \tilde{q}_n = \frac{e^{\tau/\sqrt{n}} - \left(1 + \frac{\tau}{\sqrt{n}}\right)}{e^{\tau/\sqrt{n}} - e^{-\tau/\sqrt{n}}}$$

$$M_{nt}^{(n)} = \sum_{k=1}^{nt} X_{nk}^{(n)} \quad H_{nt} = \frac{1}{2}(nt + M_{nt}) \quad T_{nt} = \frac{1}{2}(nt - M_{nt})$$

$$S_n(t) = S(0) u_n^{H_{nt}} d_n^{T_{nt}}$$

$$= S(0) \exp \left\{ \frac{\tau}{\sqrt{n}} M_{nt}^{(n)} \right\}$$

By Motion-Generating Function  $\varphi_n(u) = \mathbb{E}\left(e^{u \frac{\tau}{\sqrt{n}} M_{nt}^{(n)}}\right)$

$$\lim_{n \rightarrow \infty} \varphi_n(u) = \exp \left\{ u(1 + \frac{1}{2} u^2 \tau^2) \right\}$$

$\downarrow$   
 $\sigma^2 t$   
 $(r - \frac{1}{2} \tau^2)t$

$$S(t) = S(0) \exp \left\{ \tau W + \left(r - \frac{1}{2} \tau^2\right) t \right\}$$

Brownian Motion

$$[WW](t) = t$$

$$\mathbb{E}[W_t | \mathcal{F}_s] = W(s)$$

$$\text{Var}(W(t) - W(s)) = t-s$$

Geometric Brownian Motion

$$S(t) = S(0) \exp \left\{ \tau W - \frac{1}{2} (\alpha - \frac{\sigma^2}{2}) t \right\}$$

$$\mathbb{E}[S_t] = \exp \left\{ \alpha t - \frac{1}{2} \sigma^2 t \right\}$$

$$\mathbb{E}[Z(t) | \mathcal{F}_s] = Z(s)$$

$Ito$  process

$$X = \int \sigma dW + \int \left( \alpha - \frac{1}{2} \sigma^2 \right) dt$$

Percentage drift  
Percentage volatility  
 $Ito$ -integral  
volatility  
instantaneous rate of return

$$S(t) = \text{geometric Brownian} = e^X S(0) \Rightarrow \int dS(t) = \int \alpha S(t) dt + \int S(t) \sigma dW$$

Solution

$$S(t) = S(0) \exp \left\{ \tau W + \left( \alpha - \frac{1}{2} \sigma^2 \right) t \right\}$$

$$\mathbb{E}[S(t)] = e^{\alpha t} S(0) \quad \text{Mean rate of return is } \alpha$$

Analysis :  $Ito$  integral  $I = \int_0^T \Delta(u) dW(u)$   $\Delta(t)$  is an adapted stochastic process

① continuity    ② Adaptivity    ③ Linearity  
④ Martingale    ⑤  $Ito$  Isometry    ⑥ Quadratic variation

$$\mathbb{E}[I | \mathcal{F}_s] = I_s \quad \mathbb{E}[I^2] = \mathbb{E} \int_0^t \Delta^2 du \quad [I, I]^{(t)} = \int_0^t \Delta^2 du$$

2

Why the price calculated by Binomial tree and Black-Scholes Model are (12) convergent? (Their mean square error convergent to 0)

To figure out why, I make an analogy in this question using Stochastic Calculus knowledge.

1. Our Assumption = Risk-neutral measure

In a complete market, a derivative's price is the discounted expected value of the future payoff under the unique risk-neutral measure.

2. Binomial tree model =

$$u = \exp\left\{\frac{\sigma\sqrt{t}}{\sqrt{n}}\right\} \quad d = \exp\left\{-\frac{\sigma\sqrt{t}}{\sqrt{n}}\right\} \quad (2.1)$$

$$\text{Using risk-neutral assumption} = P_u = \frac{e^{\alpha(\sqrt{t}/n)} - d}{u - d} \quad P_d = \frac{u - e^{\alpha(\sqrt{t}/n)}}{u - d} \quad (2.2)$$

$$\text{Stock Price} \quad S(t) = S(0) \cdot u^{H(t)} \cdot d^{T(t)} \quad (2.3) \quad H = \text{number of "up"} \quad T = \text{number of "down"}$$

$$\text{(call) Option Price} \quad f_{call}^{(n)}(t) = [P_u f_u^{(n)}(t_{k+1}) + P_d f_d^{(n)}(t_{k+1})] \quad f^{(n)}(t) = [S(t) - K]^+ \quad (2.4)$$

3. Black-Scholes-Merton Model =

1) Portfolio Value = Capital Gain + Interest Earnings on cash position

a) Capital Gain =  $\int \Delta dS$  in which  $\Delta$  is the shares of Stock

$$\text{Stock Price} = S(t) = \int_0^t \alpha S(t) dt + \int_0^t S(t) dW \text{ or } dS = \alpha S dt + \sigma S dW \quad (3.1)$$

$$b) \text{Interest Earnings} = r(X - \Delta S) dt \quad (3.2)$$

$$\text{Combine a) b)} : dX(t) = r X(t) dt + (\alpha - r) \Delta S(t) dt + \sigma \Delta S(t) dW \quad (3.3)$$

3

Call

$$2) \text{ Option Value } c = (S(T) - K)^+$$

Assumption : Option Value is only the function depends on time  
and the value of the stock price at that time

Right? → (It can be automatically inherit from risk-neutral assumption)

$$c = c(t, S_t)$$

3) Equating the Evolution (the risk-neutral assumption)

$$e^{-rt} X^{(t)} = e^{-rt} c(t, S_t) \quad (3.4)$$

$$d(e^{-rt} X^{(t)}) = d[e^{-rt} c(t, S_t)] \quad (\text{BSM equation}) \quad (3.5)$$

With boundary condition, we get the solution =

$$c(t, x) = x N(d_+) - K e^{-r(T-t)} N(d_-) \quad (3.6)$$

$$\text{When } t=T, \quad c(T, x) = (x - K)^+ \quad (3.7)$$

Stock Price

Analysis = the formula above from 1, 2, 3 are used in our price modeling  
Let us figure out what would happen when we increase the "steps n"

1) Stock Price  $S_t$  From (2.3) combined with (2.1)

$$\begin{aligned} S(t) &= S(0) \cdot e^{H^{(n)}(t)} \cdot e^{T^{(n)}(t)} \\ &= S(0) \exp \left\{ \frac{D\bar{T}}{n} \cdot H^{(n)}(t) \right\} \exp \left\{ \frac{-D\bar{T}}{n} \cdot T^{(n)}(t) \right\} \end{aligned} \quad (4.1)$$

Moreover the Symmetric Random Walk  $M_K = \sum_{j=1}^K X_j$   $X_j = \begin{cases} 1 & \text{if } H \\ -1 & \text{if } T \end{cases}$

We have  $n = H + T$

$M = H - T$  which is a random walk, a stochastic process

$$\begin{cases} H^{(n)}(t) = \frac{1}{n} (M^{(n)} + n) \\ T^{(n)}(t) = \frac{1}{n} (M^{(n)} - n) \end{cases} \quad (4.2)$$

4

Substitute (4.2) into (4.1), we get

$$S^{(n)}_t = S^{(0)} \exp \left\{ \frac{\sigma^2 t}{2n} M^{(n)}_t \right\} \quad (4.3)$$

As  $P_u$  and  $P_d$  are not equal to  $\frac{1}{2}$ , we cannot make the equation that  $M^{(n)}_t/n = W^{(n)}_t$  which is a random Walk.

Rather than making  
So, Besides the effort to make similarity, I calculate the  
moment-generating function of  $S^{(n)}_t$

$$\psi^{(n)}(u) = \mathbb{E}[e^{uS^{(n)}_t}] = \mathbb{E}\left[\exp\left\{u \cdot S^{(0)} \exp\left\{\frac{\sigma^2 t}{2n} M^{(n)}_t\right\}\right\}\right] \quad (4.4)$$

$$\lim_{n \rightarrow \infty} \psi^{(n)}(u) = \exp \left\{ u \left[ (r - \frac{1}{2}\sigma^2)t \right] + \frac{1}{2}u^2(\sigma^2 t) \right\} \quad (4.5)$$

So as  $n \rightarrow \infty$ ,  $S^{(n)}_t$  is log Normal  $\sim N(r - \frac{1}{2}\sigma^2)t, \sigma\sqrt{t}$

$$\text{Moreover } S_t = \lim_{n \rightarrow \infty} S^{(n)}_t = S^{(0)} \exp \left\{ rW + (r - \frac{1}{2}\sigma^2)t \right\} \quad (4.6)$$

Using Ito-Doeblin Formula =

$$dS_t = S_t \sigma dW + rS_t dt \quad (4.7)$$

Comparing (4.7) with what we used in Black-Scholes Model

The Binomial tree we used is the Special Case, that  $\alpha = r$

As we discussed above, in BSM model Option price ~~is no~~ is irrelevant (uncorrelated) with  $\alpha$ . It can also be concluded from Risk-Neutral Assumption.

Moreover, focus on Option Pricing

### 5. Option Price Analysis

- Option Pricing
- 1) In Binomial tree we used the risk-neutral Assumption that  
 $f(t_k) = e^{\frac{rt}{\sigma^2}} [P_u f_u(t_{k+1}) + P_d f_d(t_{k+1})] \dots (2.4)$
  - 2) Boundary Condition  $f(T) = (S(T) - K)^+$
- 1) In Black-Scholes-Merton, we also used the risk-neutral Assumption
- $d(e^{-rt} X) = d(e^{-rt} c(t, S(t))) \dots (3.5)$
  - 2) Boundary Condition  $c(T, S(T)) = (S(T) - K)^+$

### b. Conclusion.

- 1) The stock price in both models ~~are~~ <sup>convergent to or submit to</sup> geometric Brownian Motion
- $$S(t) = S_0 \exp \left\{ rW + \left( r - \frac{\sigma^2}{2} \right)t \right\}$$
- $$dS(t) = \alpha S(t) dt + \sigma S(t) dW$$
- 2) The option price in both models ~~due to~~ <sup>due to</sup> risk-neutral theory and ~~same~~ has the same boundary condition.

Hence, it's not that surprised that ~~why~~ price estimated by Binomial tree and BSM model are so equal as  $n$  increased.