# **Computational pricing of options**

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# Organization

# Anderson localization in more than one particle systems

- What is Anderson Localization?
- One particle localization illustrate by fractional moment
- multiparticle/many body localization
- Introduction to Holstein model distinguished particle with many Bosons
  - Intermediate multiparticle to many body model

# **Scaling Binomial model**

- Assume initial security price  $S_0$  with volatility  $\sigma$ .
- Assume risk-free bond with yearly effective interest rate re
- Divide year into N steps
- The step-wise interest becomes:

$$r\delta = (1+r_e)^{1/N} - 1 \approx r_e/N$$

The value of a bond is:  $A_{\frac{k}{N}} = (1 + r\delta)^{\frac{k}{N}}$ 

The stepwise fluctuations are

$$\sigma \Delta = \sigma / \sqrt{N}$$

• le  $\Delta^2 \approx \delta$ 

• The upwards and downwards fluctuations are,

 $m_{\pm} = r\delta \pm \sigma\Delta$ 

• The values of the stock price are

$$S_0, S_{1/N}, S_{2/N}, \dots$$

so that

$$\mathbb{P}(\mathcal{S}_{\frac{k+1}{N}} = (1+m_+)\mathcal{S}_{\frac{k}{N}}) = \mathbb{P}(\mathcal{S}_{\frac{k+1}{N}} = (1+m_-)\mathcal{S}_{\frac{k}{N}}) = rac{1}{2}$$

4

#### **Cox - Rubinstein - Ross**

- Consider a European call with payoff H = (S<sub>T</sub> − X)<sup>+</sup> at time of expiry T, in discrete steps the expiry time is step n = ⌊NT⌋
- Lower integration bound  $S_T > X$ :

$$k_0 = \left[ rac{\ln rac{X}{(1+m_-)^n S_0}}{\ln rac{1+m_+}{1+m_-}} 
ight]$$

Value:

$$V_0 = \sum_{k=k_0}^n \frac{1}{2^n} \binom{n}{k} \left[ S_0 \left[ \frac{1+m_+}{1+\Delta_r} \right]^k \left[ \frac{1+m_-}{1+\Delta_r} \right]^{n-k} - \frac{X}{(1+\Delta_r)^n} \right]$$

# Gaussian approximation of the CRR formula

Shifted probability: q<sub>±</sub> = ½ 1+m<sub>±</sub>/(1+rδ) = ½ 1+rδ±σΔ/(1+rδ)
Value:

$$V_0 = S_0 \mathcal{N}_+ \left(\frac{k_0 - q_+ n}{(q_+ q_- n)^{1/2}}\right) - \frac{X}{(1 + r_e)^T} \mathcal{N}_+ \left(\frac{k_0 - n_2^1}{(\frac{1}{4}n)^{1/2}}\right)$$

• Where  $\mathcal{N}_+(w) = \mathbb{P}(Z \ge w)$  for a standard normal random variable Z.

#### Approximating the approximation

Terms: as  $N \to \infty$ 

$$k_{0} \approx \frac{\ln \frac{X}{S_{0}} - n \ln(1 + m_{-})}{\ln \frac{1 + m_{+}}{1 + m_{-}}} \approx \frac{\ln \frac{X}{S_{0}} - NT(r\delta - \sigma\Delta - \frac{1}{2}\sigma^{2}\Delta^{2})}{2\sigma\Delta}$$
$$q_{+}n = \left(\frac{1 + r\delta + \sigma\Delta}{1 + r\delta}\right) \frac{NT}{2}$$
$$(q_{-}q_{+}n)^{1/2} = \frac{\left((1 + r\delta)^{2} - \sigma^{2}\Delta^{2}\right)^{1/2}}{1 + r\delta} \frac{N^{1/2}T^{1/2}}{2}$$

$$\frac{k_0 - q_+ n}{(q_- q_+ n)^{1/2}} \to \frac{\ln \frac{X}{S_0} - T(r + \frac{1}{2}\sigma^2)}{\sigma T^{1/2}} =: D_0^+$$
$$\frac{2k_0 - n}{n^{1/2}} \to \frac{\ln \frac{X}{S_0} - T(r - \frac{1}{2}\sigma^2)}{\sigma T^{1/2}} =: D_0^-$$

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Stochastic model

• Stock price model:

$$dS_t = rS_t dt + \sigma S_t dW_t$$

solution:

$$S_t = S_0 e^{(r - \frac{1}{2}\sigma^2)t + \sigma W_t}$$

- European option, payoff of  $H_T = g(S_T)$  a function of the security value at expiry.
- Value of option at time *t* if present value of security is *x*,

$$V(t,x) = e^{-r(T-t)} \mathbb{E}(g(S_T)|S_t = x)$$

**Black Scholes PDE for European Option** 

• V satisfies the PDE:

$$rV = \dot{V} + rxV' + \frac{1}{2}\sigma^2 x^2 V''$$

in x > 0 and 0 < t < T with boundary condition,

$$V(T,x) = g(x) \qquad \qquad V(t,0) = g(0)$$

#### **Black Scholes value of European Call**

- Value of Euro call at expiry:  $H_T = (S_T X)^+$ .
- Solution to the PDE:

$$V(t, S_t) = S_t \mathcal{N}_+(-D_{t,S_t}^+) - X e^{-r(T-t)} \mathcal{N}_+(-D_{t,S_t}^-)$$

where

$$D_{t,S}^{\pm} = \frac{\ln \frac{S}{X} + (r \pm \frac{1}{2}\sigma^2)(T-t)}{\sigma\sqrt{T-t}}$$

• Comparison to CRR solution, as  $N \to \infty$ :

$$V_{0} = S_{0}\mathcal{N}_{+}\left(-D_{0}^{+}
ight) - Xe^{-rT}\mathcal{N}_{+}\left(-D_{0}^{-}
ight)$$

• That is, the values converge as the step size goes to 0.

#### **Example - European call**

- Security:  $S_0 = 80$ ;  $\sigma = .03$ . Bond rate r = 1.5%
- Option T = 9 months, strike price X = 80.
- Time steps per year: N = 100.
- Value at time of issue (t = 0):

$$V_0 =$$
\$1.3497

#### Example - European call - Value at 3 months





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# **Computation - layout of program**

#### Input parameters

- Stock initial value  $S_0$  and volatility =  $\sigma$
- Risk free interest rate = r
- Option properties: Strike = X, and maturity = T
- Step size/ total number of steps = n

#### Model parameters:

- T/n amount of time per step
- (1 + rT/N) Discount factor over 1 step
- stock price fluctuation up / down per step

$$u = (1 + rT/N + \sigma\sqrt{T/N})$$
  
$$d = (1 + rT/N - \sigma\sqrt{T/N})$$

• Probability fluctuation up = Probability fluctuation down = 1/2.

• Introduce  $n + 1 \times n + 1$  matrix *S* for the values of the Stock.

• 
$$S[0,0] = S_0$$
  
for  $i = 1$  to  $n + 1$ :  $S[i,0] = S[i-1,0](1+u)$ ;  
for  $j = 1$  to  $i$ :  $S[i,j] = S[i-1,j-1](1+d)$ 

 $\therefore$   $S[i, j] \equiv$  security price at timestep *i* with i - j steps up and *j* steps down.

• Note according to the model, for t = iT/n

$$\mathbb{P}(S_t = S[i, j]) = \binom{i}{j} \frac{1}{2^i}$$

# **European Valuation**

- Introduce  $n + 1 \times n + 1$  matrix *EC* for the values of the (European Call) Option.
- Find the values of the option at expiry:
   For i = 0 to n: EC[n, i] = g(S[n, i]) = (S[n, i] X)^+
- Evolve the values from expiry back to time 0: For i = n - 1 to 0: For j = 0 to i:  $EC[i, j] = (1 + rT/n)^{-1} (\frac{1}{2}EC[i + 1, j] + \frac{1}{2}EC[i + 1, j + 1])$

# **American Valuation**

Extra step in the American (Put) Valuation, testing for early exercise.

- Introduce  $n + 1 \times n + 1$  matrix *Am* and *H* for the values of the Option and intrisic value.
- Find Intrisic values of option For i = 0 to n: For j = 0 to i: H[i, j] = (X - S[i, j])<sup>+</sup>
- Set values of Am at expiry equal to intrinsic value.
- Evolve the values from expiry back to time 0: For i = n - 1 to 0: For j = 0 to i:  $G = (1 + rT/n)^{-1} (\frac{1}{2}Am[i+1,j] + \frac{1}{2}Am[i+1,j+1])$ Am[i,j] = max(G, H[i,j])

# Greeks

The Greeks quantify the sensitivity of Options to variables in the model - ie, these are partial derivatives.

- Delta:  $\Delta \equiv \frac{\partial V}{\partial S}$ . Sensitivity with respect to security price.
- Gamma:  $\Gamma = \frac{\partial^2 V}{\partial S^2}$ . Second order sensitivity with respect to security price.
- Theta:  $\Theta \equiv \frac{\partial V}{\partial t}$ . Sensitivity with respect to time.
- Vega:  $\nu \equiv \frac{\partial V}{\partial \sigma}$ . Sensitivity with respect to security's implied volatility.

#### Delta

We can compute the Greeks directly from the results of the program:

- Delta:  $\Delta \equiv \frac{\partial V}{\partial S}$ .
- Introduce  $n + 1 \times n + 1$  matrix *Delta*

$$\Delta[i,j] = \frac{EC[i+1,j] - EC[i+1,j+1]}{S[i+1,j] - S[i+1,j+1]}$$

• According to Black Scholes model:

$$\Delta_{t,\mathcal{S}} = \mathcal{N}_+(-D^+_{t,\mathcal{S}}) = \mathcal{N}(D^+_{t,\mathcal{S}})$$

• This is the proportion of the replicating portfolio invested in the stock.

# **Example - Delta**





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# **Delta Hedging**

- Fix time step  $\eta$  for each hedging event we hedge times  $0, \eta, 2\eta, ..., n\eta = T$
- Calculate current value of portfolio V<sub>0</sub> and define initial replicating portfolio

$$V_0=X_0S_0+Y_0A_0$$

setting  $X_0 = \Delta[0, S_0]$ ;  $A_0 = 1$ ;  $Y_0 = V_0 - X_0 S_0$ 

• At time step  $(k + 1)\eta$  let  $X_{(k+1)\eta} = \Delta[(k + 1)\eta, S_{(k+1)\eta}]$  solve for  $Y_{(k+1)\eta}$ :

$$Y_{(k+1)\eta} = (X_{k\eta} - X_{(k+1)\eta})S_{(k+1)\eta} + Y_{k\eta}$$

 As the stock price varies continuously some errors occur from the discritization. Find

$$V_T - X_{(n-1)\eta}S_T - Y_{(n-1)\eta}A_T$$

#### Gamma

• Gamma: 
$$\Gamma \equiv \frac{\partial^2 V}{\partial S^2}$$
  
• Introduce  $n + 1 \times n + 1$  matrix *Gamma* (restrict to index  $[2: n - 1, 2: n - 1]$ )  
 $\Gamma[i, j] = \frac{EC[i, j + 1] - 2 * EC[i, j] + EC[i, j - 1]}{(S[i, j] - S[i, j - 1])^2}$ 

$$\Gamma_{t,S} = \frac{\phi(D_{t,S}^+)}{S\sigma\sqrt{T-t}}$$

#### **Example - Gamma**





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#### Theta

- Theta:  $\Theta \equiv \frac{\partial V}{\partial t}$
- Introduce n + 1 × n + 1 matrix Gamma (restrict to index [0 : n 2, :])

$$\Theta[i,j] = \frac{EC[i+2,j+1] - EC[i,j]}{2\delta}$$

 $\delta \equiv {\rm time \; step }$ 

• Formula due to Black-Scholes:

$$\Theta_{t,S} = -\frac{S\sigma\phi(D_{t,S}^+)}{2\sqrt{T-t}} - rXe^{-r(T-t)}\mathcal{N}(D_{t,S}^-)$$

# **Asian Option**

Asian options allow exchanges of security based on average price.
Let

$$A_T = \frac{1}{T} \int_0^T S_t dt$$

The Asian Call with fixed strike price has payoff at expiry

$$H_T = (A_T - X)^+$$

• The Asian Call with floating strike has payoff at expiry

$$H_T = (S_T - A_T)^+$$

• A mathematically simpler version of this option replaces the arithmetic average with a geometric average, that is replace  $A_T$  with

$$G_T = e^{rac{1}{T}\int_0^T \ln S_t dt}$$

#### Monte Carlo Valuation of Asian option

Note the Asian option is path dependent. Thus, for an *n* step discrete model the total number of terms/paths to average over is  $2^n$ .

This is too large to effectively implement.

Soln: Conduct trials indexed by *j*. For each *j*, simulate stock value over *n* steps:

$$A_T^{(j)} = rac{1}{n} \sum_{i=1}^n S_i^{(j)}.$$

Then approximate the value of the call by the average of results over *M* trials:

$$H_{\mathcal{T}} pprox \sum_{j=1}^M (\mathcal{A}_{\mathcal{T}}^{(j)} - \mathcal{S}_{\mathcal{T}})^+$$

#### **Example - Fixed strike Asian Call**

$$H_T = (A_T - X)^+$$

- Security:  $S_0 = 80$ ;  $\sigma = .03$ . Bond rate r = 1.5%
- Option T = 9 months, strike price X = 80.
- Time steps per year: N = 100. Total number of trials: M = 1000
- Value at time of issue (t = 0):

$$V_0 =$$
\$0.7225

• Value for geometric version: \$0.7188.

#### Example - Fixed strike Asian Call - Distribution of values

- $\mathbb{P}(H_T < .01) = \frac{362}{1000}$
- Value frequencies above 1¢

