

# **An Introduction to Computational Finance**

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## Some Facts?

- Populations in the developed world are living longer
  - More people will need to live on invested capital in their retirement years.
- Conventional wisdom
  - Stocks are good investments over the long term
- Recent study (Financial Analysts Journal, 2004)
  - An investor holding a diversified equity portfolio has a 14% chance of a negative real return over a twenty year period

## Financial Insurance

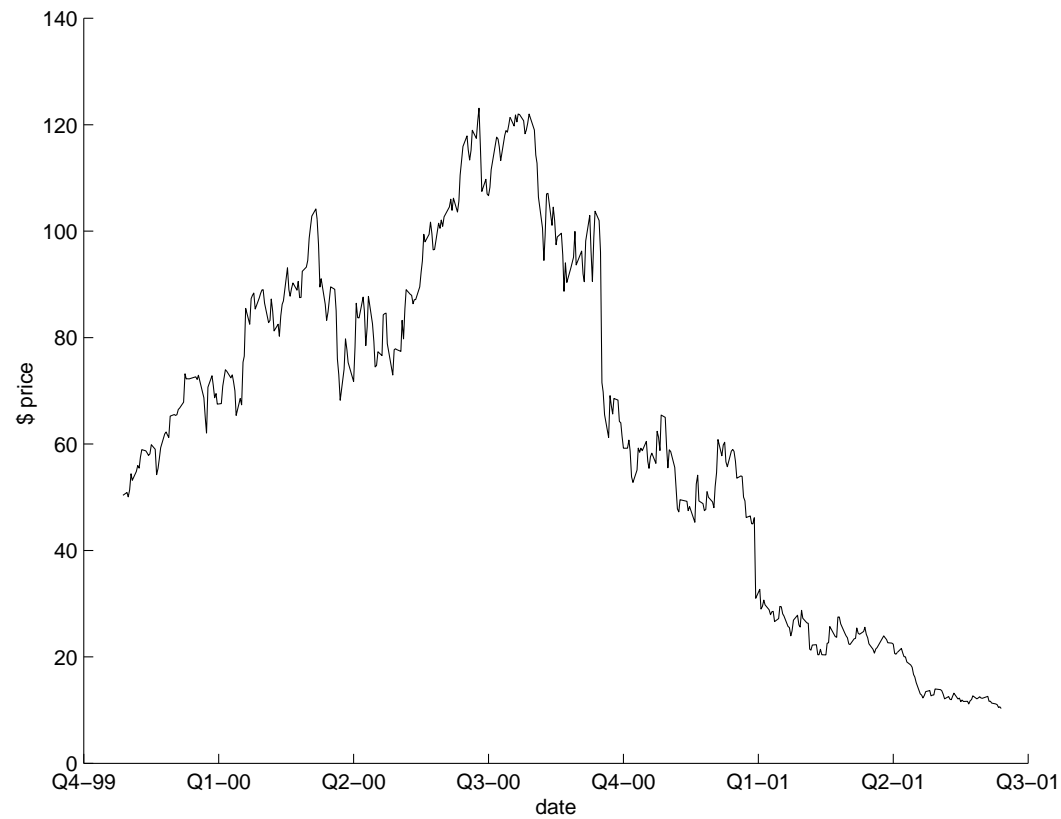
- Derivative securities (options, futures, forwards) are tools which can be used to manage risk.
- Any investment which includes some kind of protection contains an *embedded option*.
- Derivative securities are used by financial institutions to hedge risk such as
  - Currency fluctuations
  - Uncertain energy costs
  - Changes in interest rates

## Individual Investors?

- Individual investors are often unaware that they buy/sell options
- Contracts with embedded options
  - Mortgage prepayment privileges
  - Fixed rate natural gas home heating contracts
  - Equity linked GICs
- As well, many pensions plans use derivative securities in their investment portfolio

# Do we need financial insurance?

## Nortel Share Price



## Example: A call option

- Suppose I have decided that I want to buy IBM stock for  $\$K$  in 1 month's time
- But the price of IBM stock will fluctuate during the next month
- I don't want to pay more than  $\$K$  for the stock
- But if the stock falls below  $\$K$ , then I will be happy to pay less than  $\$K$  to own IBM
- Can I take out some sort of insurance to ensure that I will have to pay at most  $\$K$  for the stock?

## A call option

- A *call* option gives me the right, but *not* the obligation to purchase the stock for a specified price (the strike price  $K$  ) at some time in the future (the expiry date  $T$ ).

$S$  = value of stock

- By purchasing a call option, at strike  $K$ , expiry 1 month
  - If  $S > K$ , I exercise the option, and buy the IBM stock for  $\$K$ .
  - If  $S < \$K$ , I let the option expire, and buy the IBM stock on the open market

## The option value

- In one month, we know for sure what the option is worth
- If  $S > K$ , I can buy the stock for  $K$  and immediately sell it for  $S$
- If  $S < K$ , I will not exercise the option (why pay  $K$  for something worth  $S$ ?)
- More mathematically ( $T =$  one month)

$$\text{Value of call option} = V(T = 1 \text{ month}) = \max(S - K, 0)$$

- What is a fair price for this option today?



## A put option

Our previous example was for a *call option*

- Protection against rising prices
- Protection against falling prices can be obtained using a *put option*
- A put option is the right but not the obligation to sell an asset for the strike  $K$  at time  $T$ .
- More mathematically ( $t = T$ )

$$\text{Value of put option} = V(t = T) = \max(K - S, 0)$$

## Some Jargon

The value of the option at maturity is also called the *payoff*

$$\text{Payoff of call} = \max(S - K, 0)$$

$$\text{Payoff of put} = \max(K - S, 0)$$

A *European Option* can only be exercised at maturity  $T$ .

An *American Option* can be exercised at any time in  $[0, T]$ .

↔ The holder can decide to receive the payoff at any time.

↔ Most options traded on exchanges are American style.

We will consider a European call option in this example.

## A simple model: call option

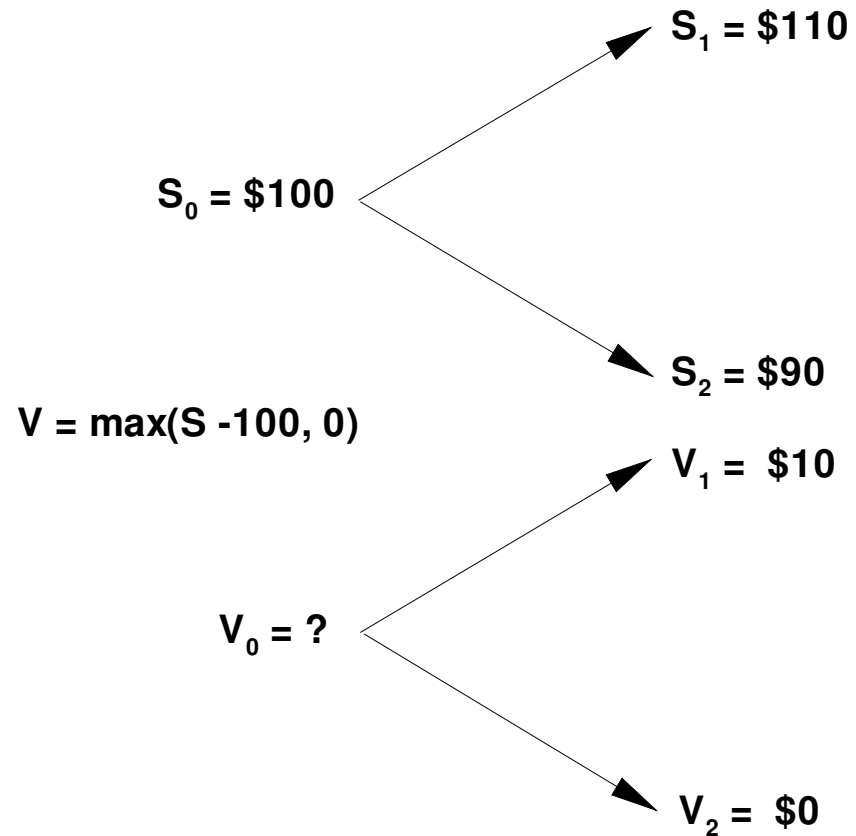
- Suppose the strike price is \$100, and the stock is trading today at  $S_0 = \$100$ .
- Let's assume a very simple model: in one month's time, the stock price can have only two possible values

$$S_0 \rightarrow S_1 = 110$$

$$S_0 \rightarrow S_2 = 90$$

## A two state tree

What is the value of the option today?  $V_0$

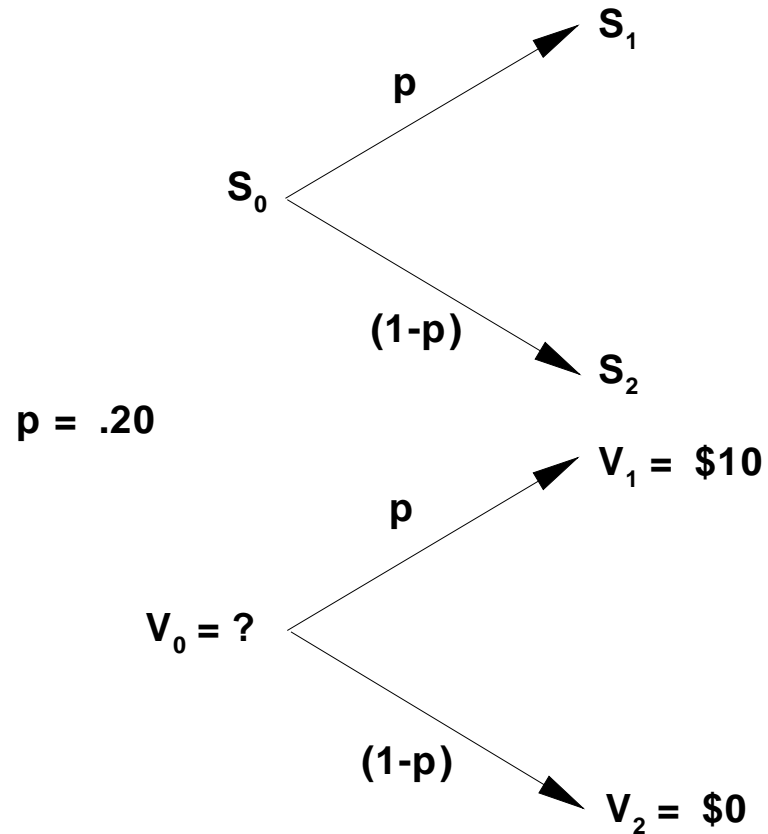


## Some extra information

Probability

$$S_0 \rightarrow S_1 = p$$

$$S_0 \rightarrow S_2 = (1 - p)$$



## Option value?

- At this point, most people would value the option today as the discounted, expected value of the payoff

$$V_0 = e^{-r\Delta t} (pV_1 + (1-p)V_2)$$

$r =$  interest rate

$\Delta t =$  One month

- To keep things simple here, let's ignore the discounting effects ( $r\Delta t \simeq 0$ )

## Option value

- This gives (in our example:  $p = .2$ )

$$\begin{aligned}V_0 &= .2 \times \$10 + .8 \times \$0 \\ &= \$2\end{aligned}$$

- Suppose I offer to buy this option from you for \$3, will you accept my offer?
- On each option, you will make an expected profit of (price - expected payout =  $3 - 2$ ) \$1

## An arbitrage opportunity

- I will be very happy to buy the option from you for \$3, since I will immediately exploit the *arbitrage opportunity*
- I can devise a trading strategy, such that I will make a profit of \$2, *regardless of whether  $S_0 \rightarrow 110$ , or  $S_0 \rightarrow 90$ .*
- How do I do this? Sounds like magic.



## The Hedging Portfolio

- To exploit this arbitrage opportunity, I will construct a portfolio  $\Pi$  which is long the option, and short  $\alpha$  shares

$$\Pi = V - \alpha S$$

- A short position means I have borrowed the security, sold it, but have to give it back at some future time.
- I will choose  $\alpha$  so that there is no uncertainty about the value of the portfolio at the expiry time of the option

## The Hedging Portfolio

- The possible portfolio values are (in 1 month)

$$\Pi_1 = V_1 - \alpha S_1$$

$$\Pi_2 = V_2 - \alpha S_2$$

- Setting  $\Pi_1 = \Pi_2$ , and solving for  $\alpha$

$$\begin{aligned}\alpha &= \frac{V_1 - V_2}{S_1 - S_2} \\ &= \frac{10 - 0}{110 - 90} = \frac{1}{2}\end{aligned}$$

## The Hedging Portfolio

- Today, I buy the option from you for \$3.  
→ I have to pay \$3.
- I borrow  $1/2$  share of the stock, and sell it ( $S_0 = \$100$ ). I will have to return this in one month's time.  
→ This gives me \$50 in cash.
- The total value of my cash is  $C_0$

$$C_0 = \$50 - \$3 = \$47$$

*Case :  $S_0 \rightarrow S_1 = 110$*

Now, in one month's time, suppose

- $S_0 \rightarrow S_1$ , which means that

$$\begin{aligned}V_1 &= \$10 \\ -\alpha S_1 &= -\frac{1}{2}(110) = -\$55\end{aligned}$$

- (I have to buy 1/2 shares at \$110, and return to broker)
- So the value of the portfolio is  $\Pi_1 = V_1 - \alpha S_1 = -\$45$ .
- But I have \$47 in cash, so I gain \$2.

$$\textit{Case} : S_0 \rightarrow S_2 = 90$$

Similarly

- $S_0 \rightarrow S_2$ , which means that

$$\begin{aligned} V_2 &= \$0 \\ -\alpha S_2 &= -\frac{1}{2}(90) = -\$45 \end{aligned}$$

- So the value of the portfolio is  $\Pi_2 = V_2 - \alpha S_2 = -\$45$ .
- But I have \$47 in cash, so I gain \$2.

## Arbitrage opportunity

- So, no matter what happens to the stock (it can go up or down, *I don't care*), I make a riskless profit of \$2.
- What would happen in reality?
  - Arbitrageurs would buy up as many options as possible
  - This would drive up the price of the option, until the option price was the *no-arbitrage* price.
- The observed market price should be the no-arbitrage price, **not** the expected payoff

## No-Arbitrage Price

- In our example, the no-arbitrage price is given by solving  $V_0 - \alpha S_0 = \Pi_1$  for  $V_0$ , giving  $V_0 = \$5$ .
- Note that we do not care what the probabilities of an up or down movement in the stock price are!
- A bank can sell me the call option for \$5(+ some profit), and construct the hedging portfolio.
- At expiry, the portfolio can be liquidated to give exactly enough to pay off the option *regardless of whether*  $S_0 \rightarrow S_1, S_2$  (plus a locked in profit)

## More Realistic Models

- Assume more complex *stochastic models* for stock prices
- But use same basic *no-arbitrage* idea
- Black-Scholes differential equation for no-arbitrage price (Scholes-Merton, Nobel Prize in Economics, 1977)
  - Can be solved using numerical algorithms for no-arbitrage price
- Note that we do not care about precise path taken by stock (we can't predict it)
- Only gross statistical properties (volatility)



## Conclusions

- Many financial products contain embedded options
- These options are *financial insurance*, which are used to minimize risk
- Even though stock prices are unpredictable
- We can determine the no-arbitrage price of an option
- We can construct a hedging strategy to payout option (insurance) no matter what happens to the stock price!
- Modern finance is now a very technical discipline (Mathematics, Statistics, Computer Science)

## More Reading

- Peter Bernstein, *Capital Ideas: the improbable origins of modern Wall street*
- Burton Malkeil, *A random walk down Wall Street*
- N. Taleb, *Fooled by Randomness*
- N. Taleb, *The Black Swan*
- *An Introduction to Computational Finance without Agonizing Pain* ([www.scicom.uwaterloo.ca/paforsyt/agon.pdf](http://www.scicom.uwaterloo.ca/paforsyt/agon.pdf))

## A Model for Stock Prices

- Observation: For every quoted price we see in the stock market: there is one buyer for every seller
- In each transaction
  - Buyer thinks price is going up
  - Seller thinks price is going down
- Conclusion: Stock prices follow a random walk (verified by statistical tests)
  - No observable patterns in prices

SDE

## Stochastic Differential Equation for Price

- Let  $S$  be the price of an underlying asset (i.e. TSX index).
- A basic model for the evolution of  $S$  through time is Geometric Brownian Motion (GBM)
- In  $t \rightarrow t + dt$ ,  $S \rightarrow S + dS$

$$\frac{dS}{S} = \mu dt + \sigma \phi \sqrt{dt}$$

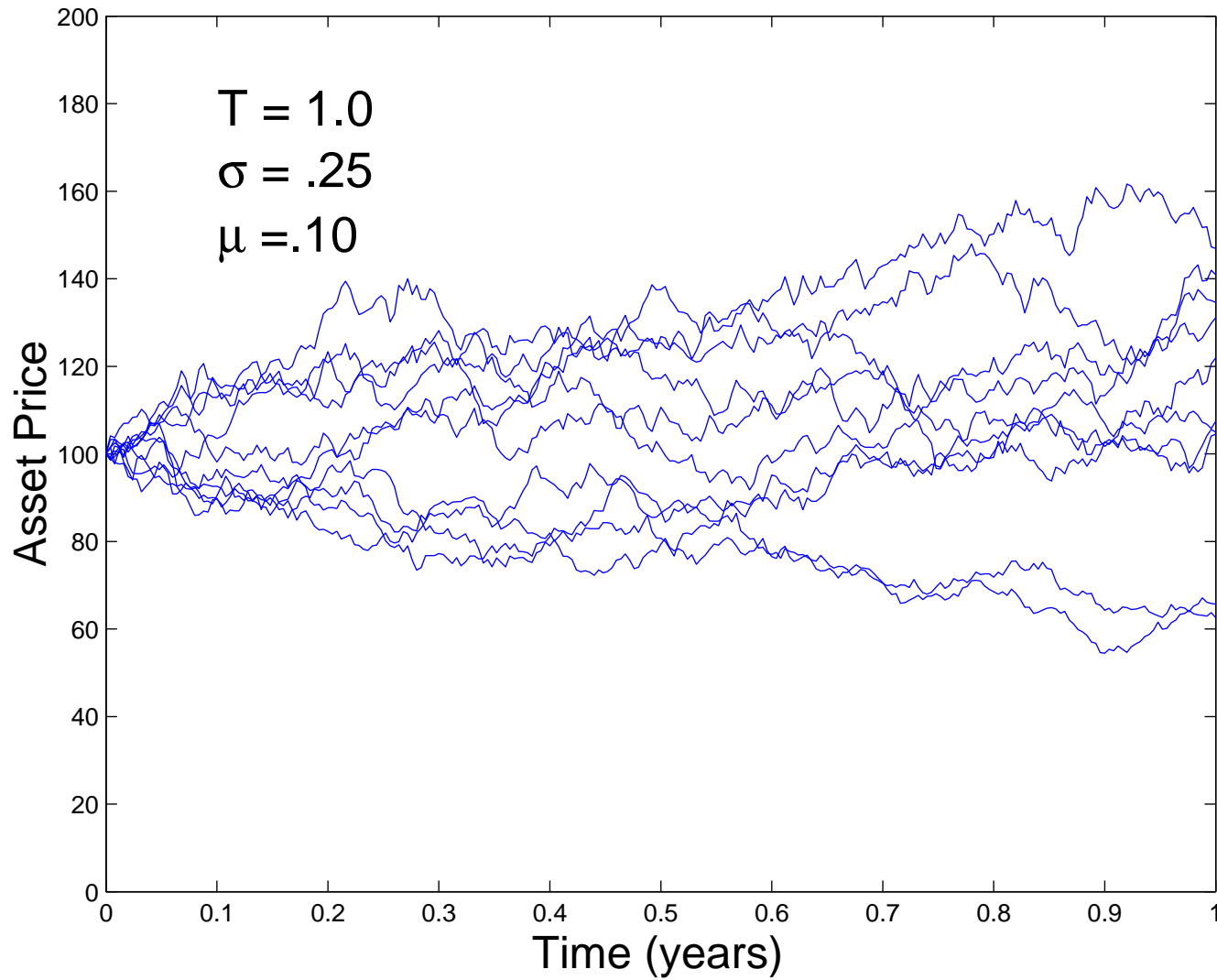
$\mu$  = drift rate,

$\sigma$  = volatility,

$\phi$  = random draw from a  
standard normal distribution

Monte Carlo Paths

## Geometric Brownian Motion



## A lattice model

In order to price an option, don't want to deal with the SDE directly.

We will develop a discrete lattice model of GBM.

Denote today's stock price ( $t = t^0$ ) by  $S_0^0$ . At  $t^1 = t^0 + \Delta t$ ,

$$S_0^0 \rightarrow S_1^1 ; \text{ up with probability } p$$

$$S_0^0 \rightarrow S_0^1 ; \text{ down with probability } q = 1 - p$$

## A lattice model

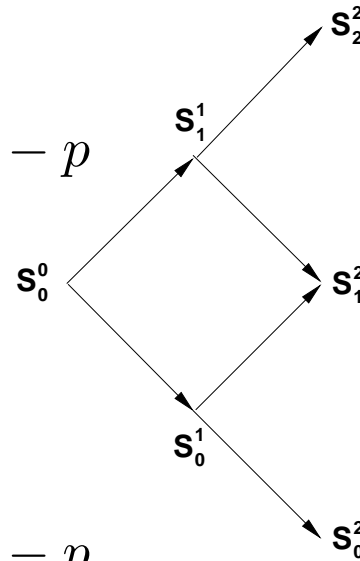
At  $t^2 = t^1 + \Delta t$ ,

$S_1^1 \rightarrow S_2^2$  ; up with probability  $p$

$S_1^1 \rightarrow S_1^2$  ; down with probability  $q = 1 - p$

$S_0^1 \rightarrow S_1^2$  ; up with probability  $p$

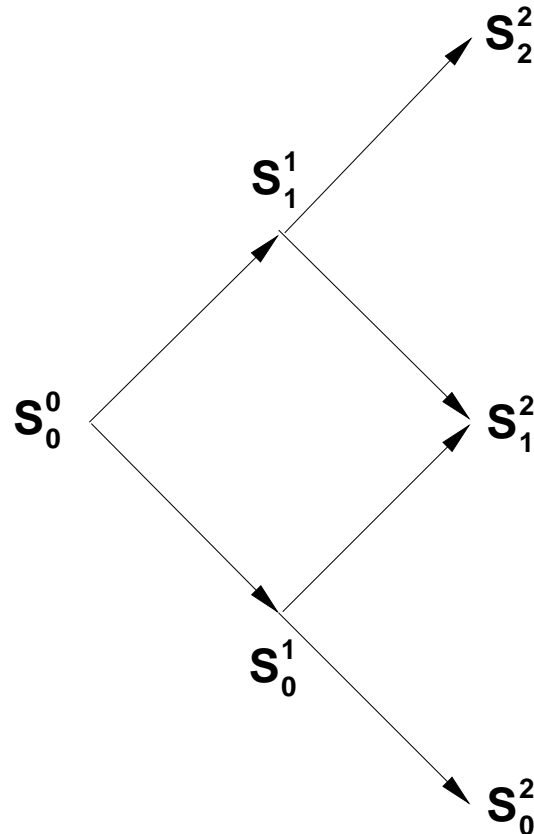
$S_0^1 \rightarrow S_0^2$  ; down with probability  $q = 1 - p$



## A Recombining Lattice

Note:  $S_j^n$   
Asset at timestep  
 $n$ ,  
node  $j$ .

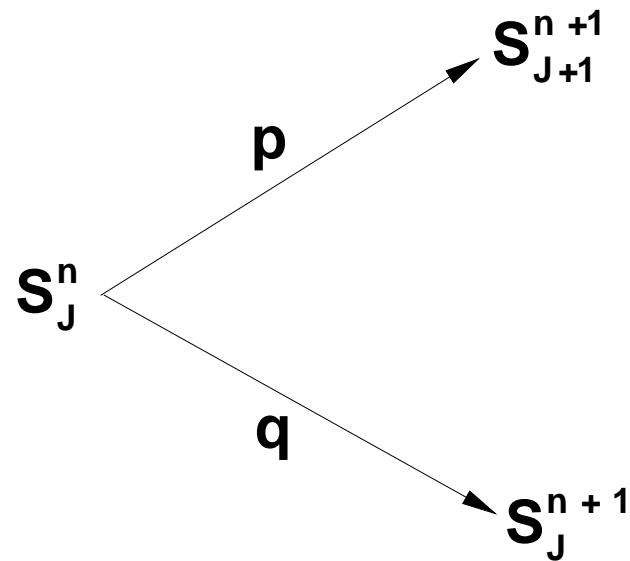
$n$  is a superscript  
not a power





## General Case

At node  $j$ , timestep  $t^n = n\Delta t$ , asset price is denoted by  $S_j^n$ .



## Consistent with GBM

If we choose:

$$S_{j+1}^{n+1} = S_j^n e^{\sigma \sqrt{\Delta t}}$$

$$S_j^{n+1} = S_j^n e^{-\sigma \sqrt{\Delta t}}$$

$$p = \frac{1}{2} \left[ 1 + \left( \frac{\mu}{\sigma} - \frac{\sigma}{2} \right) \sqrt{\Delta t} \right]$$

$$q = 1 - p$$

then, as  $\Delta t \rightarrow 0$ , random walks on this discrete lattice *converge* to the solution of the GBM SDE.

## Convergence to GBM

In other words, if we take many random walks on the lattice with these parameters, and record a histogram of the outcomes (an approximate probability density function).

Then, as  $\Delta t \rightarrow 0$ , this approximate probability density converges to the probability density function of GBM.

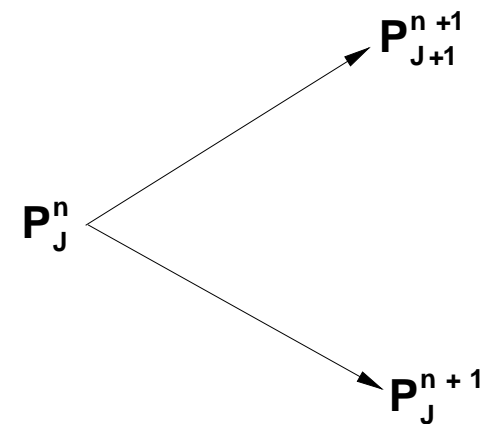
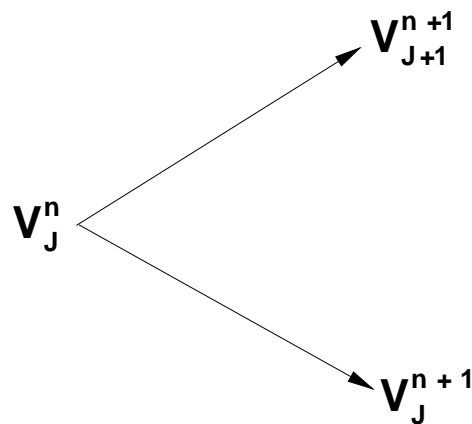
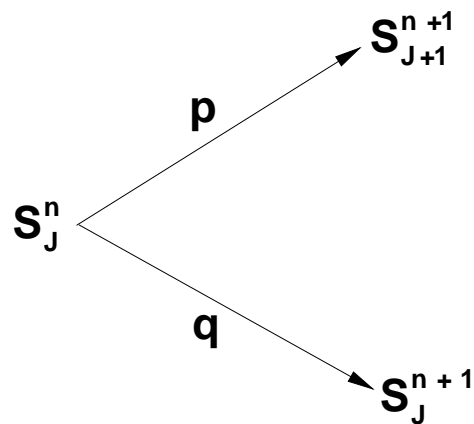
See notes for argument (not proof).

Idea: develop discrete no-arbitrage pricing model on the lattice, and then as  $\Delta t \rightarrow 0$ , this pricing model should converge to the correct solution for GBM.

## No-arbitrage Lattice

We are going to use the same idea as in our simple example. At node  $S_j^n$ , associate an option value  $V_j^n$  and a hedging portfolio

$$P_j^n = V_j^n - \alpha S_j^n$$



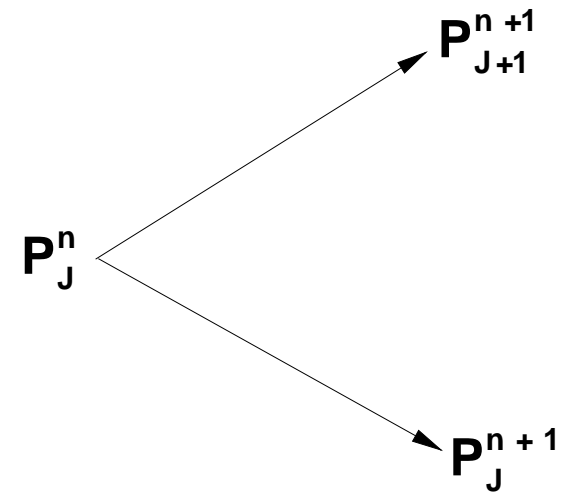
## No-arbitrage Lattice

Value of hedging portfolio at  $t = t^{n+1}$

$$P_{j+1}^{n+1} = V_{j+1}^{n+1} - \alpha S_{j+1}^{n+1}$$

$$P_j^{n+1} = V_j^{n+1} - \alpha S_j^{n+1}$$

Now, determine  $\alpha$  so that  $P_j^{n+1} =$



## No-arbitrage Lattice

$$V_{j+1}^{n+1} - \alpha S_{j+1}^{n+1} = V_j^{n+1} - \alpha S_j^{n+1} \quad (1)$$

So that

$$\alpha = \frac{V_{j+1}^{n+1} - V_j^{n+1}}{S_{j+1}^{n+1} - S_j^{n+1}} \quad (2)$$

But, this portfolio is risk free (no uncertainty about its value), so that

$$\begin{aligned} P_j^n &= e^{-r\Delta t} P_{j+1}^{n+1} \\ \rightarrow V_j^n - \alpha S_j^n &= e^{-r\Delta t} (V_{j+1}^{n+1} - \alpha S_{j+1}^{n+1}) \end{aligned} \quad (3)$$

Substitute (2) into (3)

## No-arbitrage Lattice

$$\begin{aligned} V_j^n &= e^{-r\Delta t} (p^* V_{j+1}^{n+1} + (1 - p^*) V_j^{n+1}) \\ p^* &= \frac{e^{r\Delta t} - e^{-\sigma\sqrt{\Delta t}}}{e^{\sigma\sqrt{\Delta t}} - e^{-\sigma\sqrt{\Delta t}}} \end{aligned} \quad (4)$$

Note that the real probabilities of an up/down move do not appear in (4) ( $p^*$  does not depend on the drift  $\mu$ ).

We have determined the no-arbitrage value of  $V_j^n$  in terms of  $V_{j+1}^{n+1}$ ,  $V_j^{n+1}$ .

## No-arbitrage Lattice

Recall that the no-arbitrage value is not the expected value.

But, for  $\Delta t \rightarrow 0$ , then

$$0 \leq p^* \leq 1$$

so that

$$(p^* V_{j+1}^{n+1} + (1 - p^*) V_j^{n+1})$$

looks like an expectation.

But its not the real expected value  $\rightarrow$  termed the expectation in the risk neutral world.



Lattice model

## Delta Hedging

Since our hedging portfolio is

$$P_j^n = V_j^n - \alpha S_j^n$$
$$\alpha = \frac{V_{j+1}^{n+1} - V_j^{n+1}}{S_{j+1}^{n+1} - S_j^{n+1}}$$

Note that

$$\alpha \simeq \frac{\partial V}{\partial S} = V_S$$
$$(S_{j+1}^{n+1} - S_j^{n+1}) \rightarrow 0$$

$V_S$  is called the option delta.

This hedging strategy is called *delta hedging*

## Full Lattice Algorithm

Choose  $\Delta t = T/N$ . Construct tree of prices

$$\begin{aligned} S_j^n &= S_0^0 e^{(2j-n)\sigma\sqrt{\Delta t}} \\ n &= 0, \dots, N \\ j &= 0, \dots, n \end{aligned}$$

We know the value of the option at  $t = T = t^N$

For  $j = 0, \dots, N$

$$V_j^N = \text{Payoff}(S_j^N)$$

EndFor

## Backward Recursion: European Option

```
For  $n = N - 1, \dots, 0$ 
```

```
  For  $j = 0, \dots, n$ 
```

$$V_j^n = e^{-r\Delta t} (p^* V_{j+1}^{n+1} + (1 - p^*) V_j^{n+1})$$

```
  EndFor
```

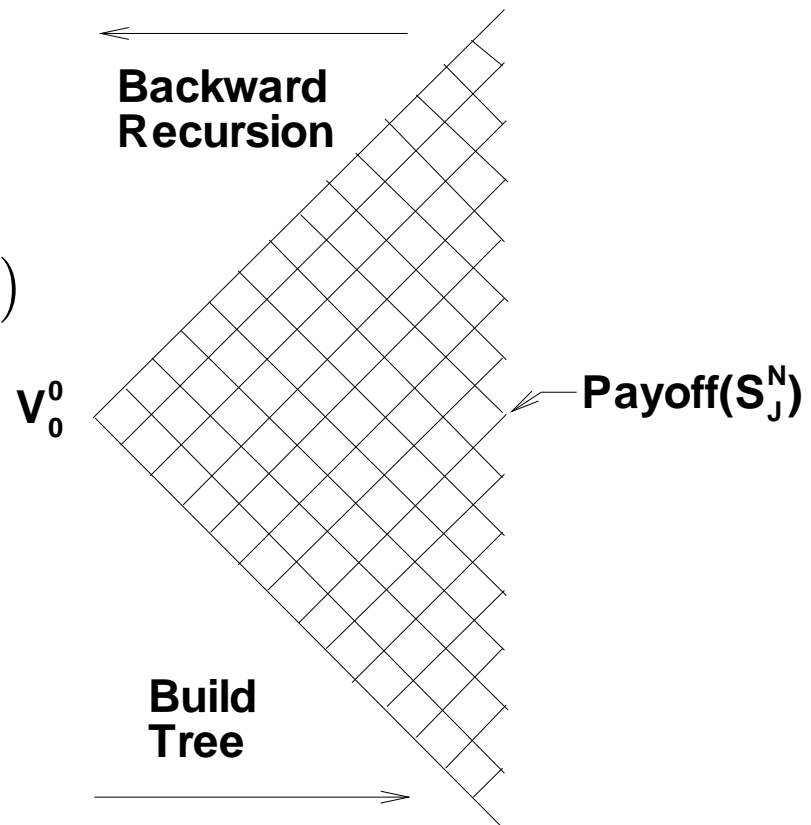
```
EndFor
```

$V_0^0$  is the no-arbitrage value of the option at  $t = t^0, S = S_0^0$ .

We also get an approximate value of the option delta =  $V_S$  at each node in the tree. This is the hedging parameter.

# Backward Recursion: European Option

$$V_j^n = e^{-r\Delta t} (p^* V_{j+1}^{n+1} + (1 - p^*) V_j^{n+1})$$



## American Options?

Recall that an American option can be exercised at any time, and the holder can receive the payoff.

So, the holder must decide, at each instant in time

- Continue to hold the option
- Exercise immediately

A rational investor will exercise if the value of exercising is larger than the value of continuing to hold.

## Backward Recursion: American Option

```
For  $n = N - 1, \dots, 0$ 
  For  $j = 0, \dots, n$ 
     $V_j^n := e^{-r\Delta t}(p^*V_{j+1}^{n+1} + (1 - p^*)V_j^{n+1})$ 
     $V_j^n := \max(V_j^n, \text{Payoff}(S_j^n))$ 
  EndFor
EndFor
```

This is a dynamic programming solution to the American option optimal exercise problem.

## Dynamic Programming

- Note that the optimal exercise of an American option requires solution of a global optimization problem
- But, since we work backwards from the end state  $N$ , we examine all possible outcomes, and choose the optimal choice at all nodes at state  $N - 1$ , and so on
- This reduces the global optimization to a set of trivial one step optimal choices

PDE

## Black-Scholes Equation

As  $\Delta t \rightarrow 0$ , the solution from the lattice algorithm converges to the solution of the Black-Scholes partial differential equation (B-S PDE)

$$\frac{\partial V}{\partial t} + \frac{\sigma^2 S^2}{2} \frac{\partial^2 V}{\partial S^2} + rS \frac{\partial V}{\partial S} - rV = 0$$

It can be shown that

$$(V^{lattice})_j^n = (V^{exact})_j^n + O(\Delta t)$$
$$\Delta t \rightarrow 0$$

↔ The lattice method is simply a numerical method for solving the B-S PDE.



Hedge

## Hedging

- Let  $V(S, t)$  be the value at any time of the option (computed from our lattice).
- The bank will sell the option to me for  $V(S, t = 0)$  today, and construct the following portfolio  $\Pi$  (*-tive*  $\rightarrow$  short)

$$\Pi = -V + \alpha S + B$$

$V$  = value of option

$S$  = price of underlying

$B$  = cash in risk free money market account

$\alpha$  = units of underlying

Note we have included a bank account  $B$  in our total portfolio  $\Pi$ .

## Hedging

So, what does the bank do?

- Sell the option today for  $V(S, t = 0)$  (lattice price).
- Construct the portfolio  $\Pi$ , by buying  $\alpha(S, t = 0)$  units at price  $S$ , and depositing  $B$  in the money market account
- As  $t \rightarrow t + \Delta t$ ,  $S \rightarrow S + \Delta S$ , bank rebalances the hedge, by buying/selling underlying so that  $\alpha(S + \Delta S, t + \Delta t) = V_S$
- Hedging portfolio is *Delta Neutral*

## Delta Hedging

- This strategy is called *Delta Hedging*
- Note that this is a dynamic strategy (rebalanced at finite intervals)
- It is self-financing, i.e. once the bank collects cash from selling option, no further injection of cash into  $\Pi$  is required.
- At time  $T$  in the future, the bank liquidates  $\Pi$ , pays off short option position, at zero gain/loss, *regardless of random path followed by  $S$ .*

## No-arbitrage Price

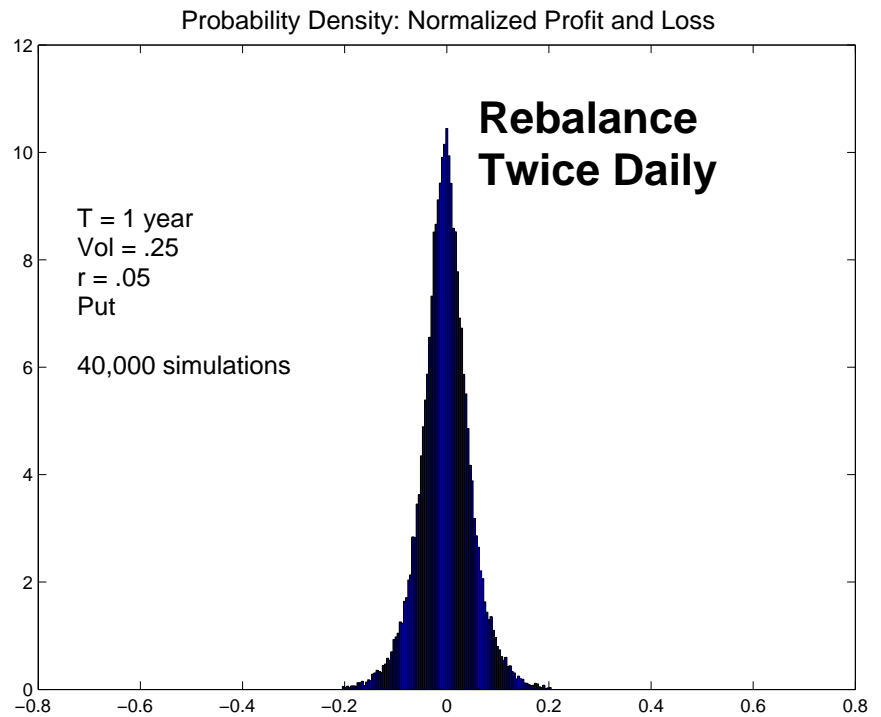
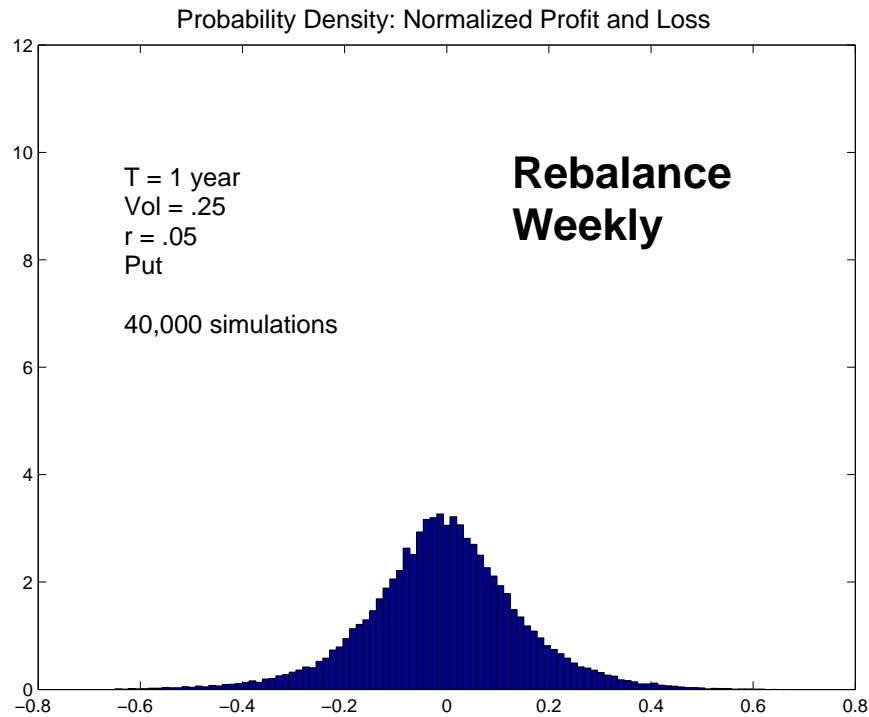
- The value of the option  $V(S, t)$  is the *no-arbitrage* value
- $V(S, t = 0)$  is the cost of setting up the portfolio  $\Pi$  at  $t = 0$
- The value of the option is *not* the discounted expected payoff

Does this actually work? Can we construct a hedge so we can't lose, regardless of the random path followed by  $S$ ?

- Simulate a random price path, along path, carry out delta hedge at finite rebalancing times (not a perfect hedge)
- Liquidate portfolio at expiry, pay off option holder, record profit and loss

Simulation

# Monte Carlo Delta Hedge Simulation: Discounted Relative Profit and Loss



## Reality

- Nobody hedges at infinitesimal intervals, volatility  $\neq \text{const.}$ , GBM not a perfect model
- Bank wants to make a profit

$$V_{buy}^{market} = V(S, t)^{model} + \epsilon_1 + \epsilon_2$$

$$V_{sell}^{market} = V(S, t)^{model} - \epsilon_1 - \epsilon_2$$

$$\epsilon_1 = \text{profit}$$

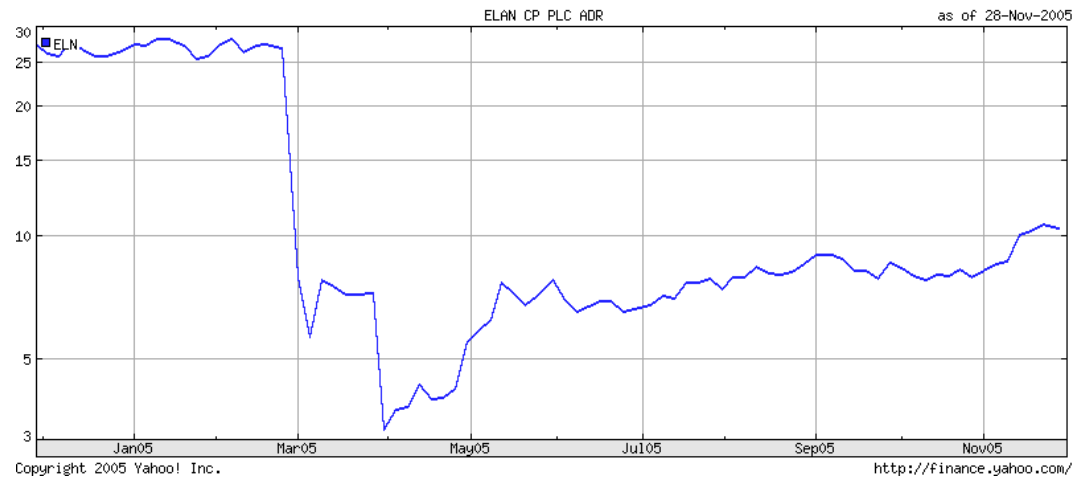
$$\epsilon_2 = \text{compensation for imperfect hedge}$$

$$V_{buy}^{market} - V_{sell}^{market} = \text{bid-ask spread}$$

GBM?

## What's Wrong with GBM?

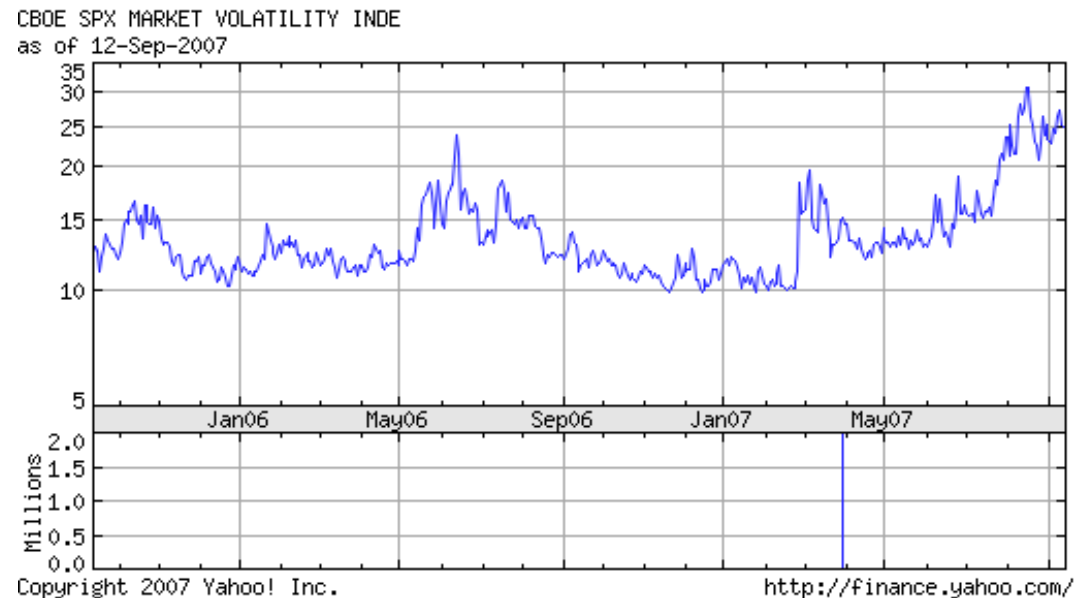
- Equity return data suggests market has *jumps* in addition to GBM
  - Sudden discontinuous changes in price



GBM?

## What's Wrong with GBM?

- Volatility not constant
- VIX index is a measure of instantaneous volatility (*S&P500*)
- Volatility is itself stochastic





## Research Challenges

- Pricing and hedging options under jump processes and stochastic volatility (Monte Carlo, PDE methods)
- Pricing exotic options (Numerical soln of PDEs)
- Optimal trade execution (algorithmic trading)
  - Optimal stochastic control
- Model calibration (optimization)

## Notes Reading

*An Introduction to Computational Finance without Agonizing Pain* ([www.scicom.uwaterloo.ca/paforsyt/agon.pdf](http://www.scicom.uwaterloo.ca/paforsyt/agon.pdf))

Sections: 1, 2.1, 2.2, 2.3, 2.4, 5

If you have time: 2.5, 2.6, 8.1, 8.2

And more if you like!

CS476, Winter 2008 *Introduction to numeric computation for financial modelling*