

Financial Derivatives#15
Pricing American and Exotic Options
via Numerical Methods

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1 Preview

- Numerical methods for option pricing
- Monte Carlo simulation
 - Formal treatment
 - Advantages and disadvantages
 - Exotic options pricing and the Greeks
- The binomial asset pricing model
 - Assumptions and model set-up
 - No-arbitrage pricing and risk neutral valuation
 - Extension to multiple periods and convergence to Black and Scholes [2]
 - The effect of dividends and the pricing of American options

2 Numerical methods for option pricing

- The beauty and convenience of the Black and Scholes [2] framework is that it is “designed” in such a manner, that it produces an analytic solution (i.e. a closed–form equation) for the prices of European calls and puts.
- However, not all option contracts (e.g. American, Asian, Barrier) admit such convenient analytic solutions. In most such cases, the PDE that the price of a non–European option must satisfy does not admit a known solution. To price such options, several numerical methods have been proposed in the literature.

- In what follows, we will demonstrate the most popular numerical methods in the literature, initially in the context of European options, and then show how to extend them for non–European contracts.
- The most popular numerical methods are:
 - Monte Carlo simulation
 - Binomial asset pricing model
 - Finite difference methods

3 Monte Carlo simulation: Formal treatment

- In 1977 Boyle [3] demonstrated how European options can be priced via Monte Carlo simulation.
- The argument is based on the *risk–neutral valuation* result, i.e.

$$c = e^{-rT} E^{\mathbb{Q}} \left[\max(S_T - K)^+ \right]$$

$$p = e^{-rT} E^{\mathbb{Q}} \left[\max(K - S_T)^+ \right]$$

- So, if we can calculate the option's expected payoff in a risk–neutral world, the option price will be given by discounting it with the risk–free rate. Assuming that interest rates are constant, this basically involves *approximating the expected risk–neutral payoff by a large number of simulated possible asset paths*. This means the following steps:

- Divide the time horizon $[0, T]$ into N , equally spaced intervals, each of length Δt .
- For each time interval, sample M , normally distributed random numbers, with zero mean and unit standard deviation ($M \times N$ random numbers in total).
- Use the random number to produce $M \times N$ matrix of possible paths for the underlying asset in a risk–neutral world. For an asset S paying a continuous dividend yield and following a gBm [in a risk–neutral world, \mathbb{Q}],

$$dS = (r - \delta) S dt + \sigma S dz$$

$$S_{t+\Delta t} = S_t \exp \left[\left(r - \delta - \frac{\sigma^2}{2} \right) \Delta t + \sigma \varepsilon (\Delta t)^{\frac{1}{2}} \right] \quad (1)$$

N.B.: Equation (1) is slightly different from the discrete version we have used in our simulation example in lecture#6 (pp. 25–26). The above is the correct and exact discrete version of a gBm, while the one in #6 is only an approximation.

- At the option maturity, $T = N \times \Delta t$, calculate the option payoff for each of the M simulated paths (i.e. a $M \times 1$ vector)

for a call, $(S_T - K)^+$ or for a put, $(K - S_T)^+$

- Calculate the average payoff over all M paths, and then discount the result with e^{-rT} .
- Boyle [3] has shown that as $M \rightarrow +\infty$ and $N \rightarrow +\infty$ (i.e. $\Delta t \rightarrow 0$), the simulated option price converges to the correct, Black and Scholes price, $O_{MC} \rightarrow O_{B\&S}$.

4 Advantages and disadvantages

- Advantages of MC simulation method:
 - Can be used when the payoff of the contract depends on the actual path followed by the underlying variable S , as well as when it depends only on the final value of S .
 - Any stochastic process for S can be accommodated, as well as any payment patterns made by the underlying asset (e.g. dividends)
 - Can be extended to accommodate situations where the option from the derivative depends on several underlying market variables.
- Disadvantages of MC simulation method:
 - It is computationally time-consuming
 - Despite some recent research results (see for example, Longstaff and Schwartz [7]), it cannot easily handle situations where there are early exercise opportunities (i.e. American, Bermudan options).

5 Calculating the Greeks via Monte Carlo simulation

- The Greek letters covered in lecture #9–10 can be calculated via Monte Carlo simulation.
- Suppose that we are interested in the partial derivative of f with respect to x , where f is the value of the option contract and x is the value of an underlying asset or a parameter that affects the option value.
- To calculate $\frac{\partial f}{\partial x}$, use MC to produce one estimate of the option price, \hat{f} as we've seen. Then assume that parameter x increases slightly by Δx , and re-calculate a new value for the option \hat{f}^* , the same way as \hat{f} (same number of time intervals N and random paths M).
- Then, an estimate of the Greek letter under review is given by

$$\frac{\hat{f}^* - \hat{f}}{\Delta x}$$

6 An exotic option example

- An *average–strike Asian put option* (denote Ap), gives its holder the right (without the corresponding obligation) to sell at time T and asset S for a price equal to the *average asset price realised* between time $t = 0$ and maturity time T , i.e.

$$Ap = \left(\frac{1}{i} \sum_{i=0}^T S_{t_i} - S_T \right)^+$$

- There is no closed–form solution for this option contract. However, we can use our Monte Carlo methodology to calculate its price.

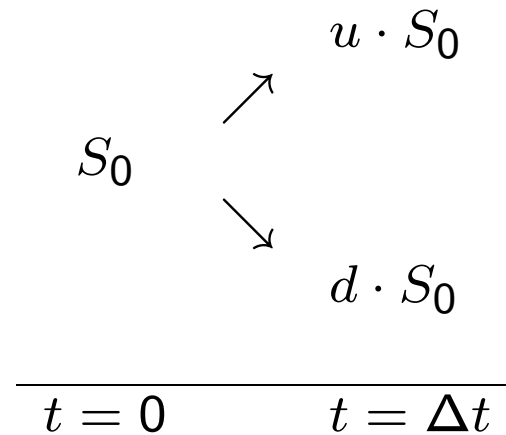
- The only difference for this contract is that the strike price is different for each asset price path: for each of the M possible path, we need to take the average over time (i.e. over N path realisations). The result is the strike price that applies for this particular path.
- Once the above is repeated for all paths M , again calculate the option's payoff, take the average payoff, and then discount the result with e^{-rT} to get the option price.

7 The Binomial Asset Pricing Model

- In 1979, Cox, Ross and Rubinstein [4] published one of the most important pedagogical papers in option pricing. Their *binomial asset pricing model* goes as follows:
- Consider the following simple market: 2 assets, 2 possible states of nature
- Assets:
 1. Underlying asset
 2. Risk free asset
- States
 1. “*u*” — Price of the underlying goes up
 2. “*d*” — Price of the underlying goes down
- If market is complete and there is to be no–arbitrage, what should be the correct price of a European call option?

8 Underlying asset

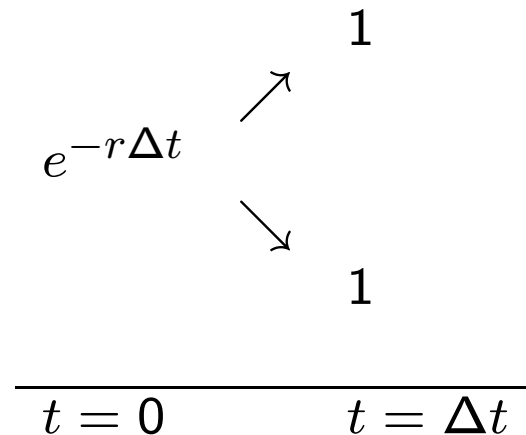
- The price of the underlying asset follows the *binomial process*:



where S_0 is the price of the asset in the beginning of the period. The price at $t = \Delta t$ will be either $S_u = u \cdot S_0$ or $S_d = d \cdot S_0$.

9 Risk free asset

- The riskless asset pays $\pounds 1$ at $t = \Delta t$ no matter which state of nature prevails. Its price follows



- This asset is also referred to as *pure discount* or *zero-coupon bond* (with “normalised” face value of $\pounds 1$). If you buy a zero-coupon bond for $\pounds e^{-r\Delta t}$ today, you will get $e^{-r\Delta t}e^{r\Delta t} = 1$ in a period’s time.

- Assumption: constant, continuously compounded riskless rate.
- Parameters d , u and r verify the following relation:

$$0 < d < 1 < e^{r\Delta t} < u \quad (2)$$

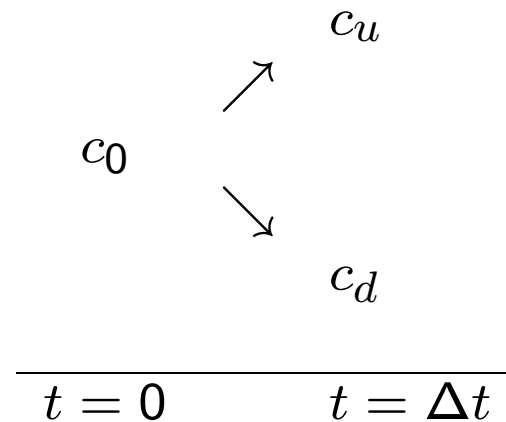
- If $e^{r\Delta t} \geq u$ no one would invest in the stock since the rate of the riskless investment is at least as great, and sometimes greater than the return of the stock.
- If $d \geq e^{r\Delta t}$ one could borrow money and invest in the stock since, even in the worst case, the stock price rises at least as fast as the debt used to buy it.
- Sometimes one assumes $d = \frac{1}{u}$ to simplify results.

10 Isn't this too simple?

- In reality, stock price movements are far more complicated than those implied by the binomial asset pricing model.
- However, this simple model is useful because:
 1. Within the binomial model the concept of *no–arbitrage pricing* and its relation to *risk–neutral* pricing is clearly illuminated.
 2. The model is used widely in practice because with a sufficient number of time–steps, it provides a good, *computationally tractable* approximation to continuous–time models (which is the benchmark in the literature).
 3. The method can be extended to handle cases that the (continuous–time) B&S cannot.

11 Arbitrage pricing

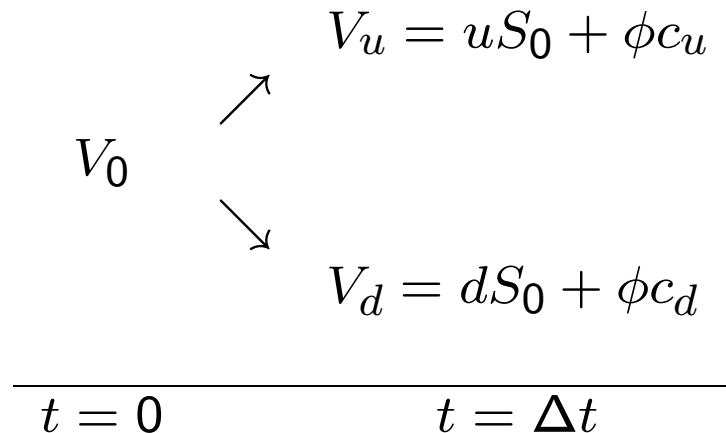
- Depending on which state prevails at time $t = \Delta t$, a European call option will have different values denoted c_u and c_d



- To price the option, at time $t = 0$ we construct an *no–arbitrage portfolio* containing:
 - One underlying asset S_0
 - ϕ calls c_0
- The value of the portfolio is

$$V_0 = S_0 + \phi c_0$$

At $t = \Delta t$ we will have



- Now choose ϕ in such a manner, so that the no–arbitrage portfolio becomes *riskless*, i.e. it yields a risk free rate of return $e^{r\Delta t}$.

Therefore, the no–arbitrage condition is:

$$V_u = V_d = V_0 e^{r\Delta t} \quad (3)$$

or equivalently

$$uS_0 + \phi c_u = dS_0 + \phi c_d = (S_0 + \phi c_0) e^{r\Delta t}$$

Basic algebra yields the solution

$$\phi = \frac{S_0 (u - d)}{c_d - c_u} \begin{cases} c_u \geq c_d \\ \leq \end{cases} 0 \quad (4)$$

and

$$c_0 = e^{-r\Delta t} \left\{ \frac{e^{r\Delta t} - d}{u - d} c_u + \left[1 - \frac{e^{r\Delta t} - d}{u - d} \right] c_d \right\} \quad (5)$$

- The coefficient ϕ is interpreted as *the number of call options necessary to include in an arbitrage portfolio containing one asset*.
- The negative sign of ϕ implies that we *sell or write* call options against the asset.
- In other words, ϕ is *the number of call options that one can write against the asset, so that a portfolio containing these and one asset is risk-free (perfect hedge)*.
- The same analysis could be done if we *write one call and we want to know how many assets we must be long*. In this case, take the inverse of ϕ and reverse its sign. This is known as the *hedge ratio* or “*delta*” of a call.

$$\Delta = -\frac{1}{\phi} = \frac{c_u - c_d}{S_u - S_d} \geq 0 \quad (6)$$

12 Risk–neutral probability

- In the no–arbitrage result for c_0 in (5), define

$$\mathbb{Q} \equiv \frac{e^{r\Delta t} - d}{u - d} \quad (7)$$

as the *risk–neutral* probability.

- Indeed, \mathbb{Q} has the necessary properties of a probability measure. Property (2) ensures that

$$0 < \mathbb{Q} < 1$$

- The amazing thing about the pricing formula in (5) is that we do not need the *actual probabilities** of states “ u ” and “ d ”, say \mathbb{P} and $1 - \mathbb{P}$, to price options.

*Also known as: historical, objective, subjective, empirical or standard.

13 Binomial formula

- The price of a European option is equal to its expected value computed using risk–neutral probabilities \mathbb{Q} , $1 - \mathbb{Q}$ and then discounted using the riskless rate r .
- From (5) using notation (7) we get:

$$c_0 = e^{-r\Delta t} [\mathbb{Q} \cdot c_u + (1 - \mathbb{Q}) \cdot c_d] = e^{-r\Delta t} \mathbb{E}^{\mathbb{Q}} [\tilde{c}_1 | \mathcal{F}_0] \quad (8)$$

where \tilde{c}_1 denotes the random price of the option which could be either $c_1 = c_u$ or $c_1 = c_d$.

- The expectation taken using risk–neutral probabilities \mathbb{Q} , $1 - \mathbb{Q}$ is denoted $\mathbb{E}^{\mathbb{Q}} [.]$ and the set $\{\mathbb{Q}, 1 - \mathbb{Q}\}$ is called risk–neutral probability measure or \mathbb{Q} –measure. The expectation taken is conditional on the information \mathcal{F}_0 available at time $t = 0$.

14 Multiperiod setting

- Let $t = 0, \Delta t, 2\Delta t, \dots, n \cdot \Delta t, \dots T$ where $T = N \cdot \Delta t$
- It is straightforward to generalise (8) to multiperiod binomial trees
- On each sub–tree

$$c_n = e^{-r\Delta t} \left[\mathbb{Q} c_{n+1}^{(u)} + (1 - \mathbb{Q}) c_{n+1}^{(d)} \right] = e^{-r\Delta t} \mathbb{E}^{\mathbb{Q}} [\tilde{c}_{n+1} | \mathcal{F}_n] \quad (9)$$

Use the above as many times as there are tree nodes between $t = 0$ and T (N times)

$$\begin{aligned} c_0 &= e^{-r\Delta t} \mathbb{E}^{\mathbb{Q}} [\tilde{c}_1 | \mathcal{F}_0] \\ &= e^{-r2\Delta t} \mathbb{E}^{\mathbb{Q}} \left[\mathbb{E}^{\mathbb{Q}} [\tilde{c}_2 | \mathcal{F}_1] | \mathcal{F}_0 \right] \\ &= \dots \\ &= e^{-rT} \mathbb{E}^{\mathbb{Q}} \left[\mathbb{E}^{\mathbb{Q}} \left[\dots \mathbb{E}^{\mathbb{Q}} [\tilde{c}_T | \mathcal{F}_{T-\Delta t}] | \dots | \mathcal{F}_1 \right] | \mathcal{F}_0 \right] \end{aligned}$$

- But from the properties of conditional expectations we know that[†]

$$\mathbb{E}^{\mathbb{Q}} \left[\mathbb{E}^{\mathbb{Q}} [\tilde{c}_{n+2} | \mathcal{F}_{n+1}] | \mathcal{F}_n \right] = \mathbb{E}^{\mathbb{Q}} [\tilde{c}_{n+2} | \mathcal{F}_n]$$

Pursuing this further yields

$$c_0 = e^{-rT} \mathbb{E}^{\mathbb{Q}} [\tilde{c}_T | \mathcal{F}_0]$$

However, we know that at maturity, a call option is

$$\tilde{c}_T = (\tilde{S}_T - K)^+$$

[†]This property is referred to as the “tower law”.

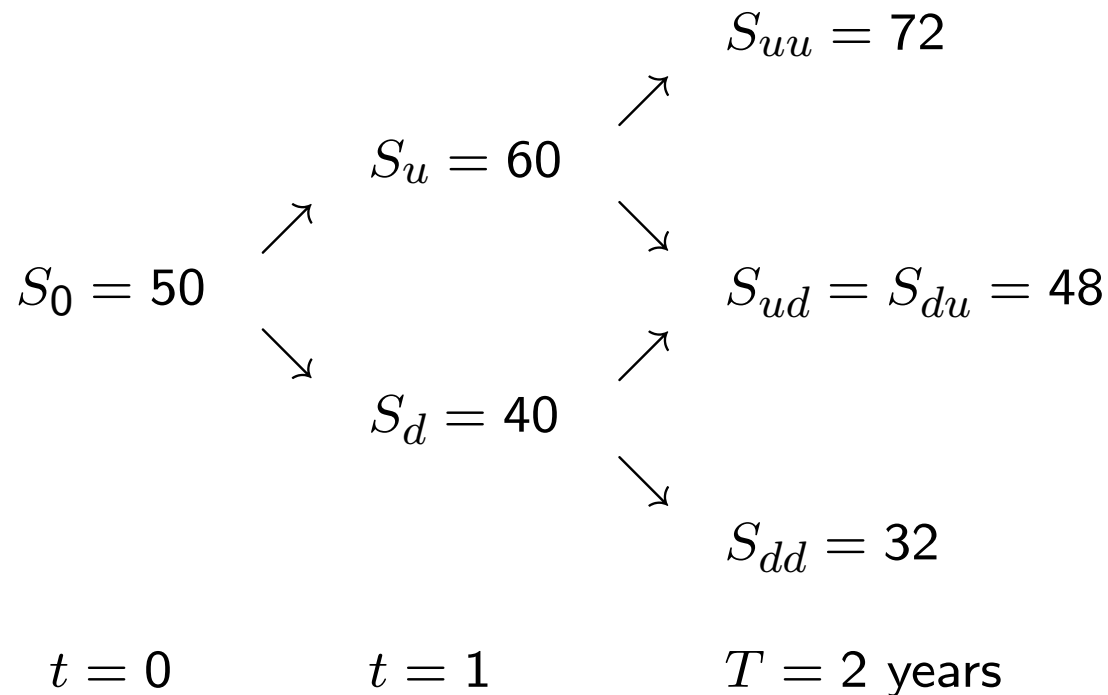
15 European option pricing formula

$$c_0 = e^{-rT} \mathbb{E}^{\mathbb{Q}} \left[\left(\tilde{S}_T - K \right)^+ \mid \mathcal{F}_0 \right] \quad (10)$$

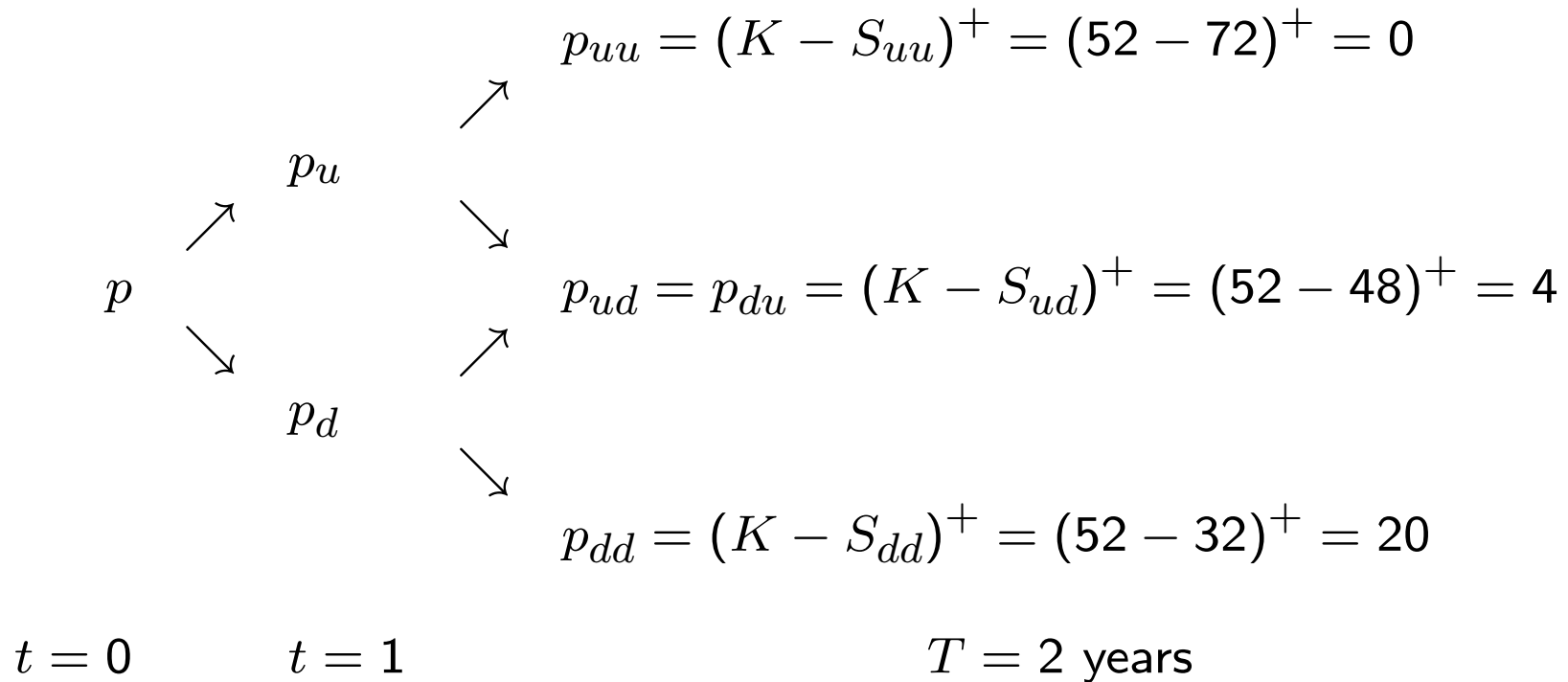
The value of a European call is a (conditional) expectation — under the risk-neutral probability measure—of its payoff at maturity, discounted using the riskless interest rate. This is the exact same result as equation (11) in lecture #4.

16 A two–period put option example

You observe a European put written on a stock with $S_0 = \text{£}50$, $K = \text{£}52$, $r = 5\%$ p.a. and $T = 2$ years. Let $\Delta t = 1$ year, $u = 1.20$ and $d = 0.80$ so that



Use a binomial tree to price the option.



SOLUTION:

1. Calculate the payoff at all final tree nodes ($T = 2$).
2. Calculate the risk–neutral probabilities $\mathbb{Q} = \frac{e^{0.05 \times 1} - 0.8}{1.2 - 0.8} = 0.6282$ and $1 - \mathbb{Q} = 0.3718$
3. Use equation (10): Multiply the payoff of each ending node with the risk–neutral probability of reaching this node, discount with the risk–free rate to time $t = 0$ and sum

$$\begin{aligned} p &= e^{-0.05 \times 2} \left[0.6282^2 \times 0 + 2 \times 0.6282 \times 0.3718 \times 4 + 0.3718^2 \times 20 \right] \\ &= 4.1923 \end{aligned}$$

17 From the Binomial model to Black–Scholes

- We will demonstrate how the Binomial model can be extended to continuous–time and how it relates to the Black and Scholes [2] formula
- A binomial tree with N steps (each of size Δt , such that $T = N \cdot \Delta t$) has $N + 1$ end–nodes and 2^N possible paths of reaching the end–nodes.
- The risk–neutral probability of reaching a particular end–note is given by

$$\mathcal{B}(n|N, \mathbb{Q}) = \frac{N!}{(N-n)!n!} \mathbb{Q}^n (1 - \mathbb{Q})^{N-n}$$

while the call option payoff is

$$\left(u^n d^{N-n} S_0 - K \right)^+$$

where $n = 0, 1, 2, \dots, N$ is the number of upwards movements in the un-

derlying asset.[‡] Thus

$$c = e^{-rT} \sum_{n=0}^N \frac{N!}{(N-n)!n!} \mathbb{Q}^n (1 - \mathbb{Q})^{N-n} \left(u^n d^{N-n} S_0 - K \right)^+$$

- Separate the two elements $u^n d^{N-n} S_0$ and K in the max operator by only considering the paths that end up with a positive payoff

$$c = S_0 \sum_{n=a}^N \frac{N!}{(N-n)!n!} (\mathbb{Q}')^n (1 - \mathbb{Q}')^{N-n} - K e^{-rT} \sum_{n=a}^N \frac{N!}{(N-n)!n!} \mathbb{Q}^n (1 - \mathbb{Q})^{N-n}$$

[‡]Remember, $\forall x \in \mathbb{Q}$, $x! = 1 \cdot 2 \cdot \dots \cdot x$, and $0! \equiv 1$.

$$c = S_0 \mathcal{B}(n \geq a | N, \mathbb{Q}') - K e^{-rT} \mathcal{B}(n \geq a | N, \mathbb{Q}) \quad (11)$$

where $\mathcal{B}(n \geq a | N, \mathbb{Q}')$ is the complementary binomial distribution function, a is the smallest non-negative integer greater than

$$a > \frac{\ln(K/S_0 d^n)}{\ln(u/d)}$$

and

$$\mathbb{Q}' = \frac{u}{e^{rT}} \mathbb{Q}$$

- Cox, Ross and Rubinstein [4] show that if the u and d parameters are set so as to match the volatility σ of the Black–Scholes framework, i.e.

$$u = e^{\sigma(T/N)^{0.5}}$$

$$d = e^{-\sigma(T/N)^{0.5}}$$

then equation (11) converges to the Black and Scholes [2] for a large number of time steps N since

$$\lim_{N \rightarrow +\infty} \mathcal{B}(n \geq a | N, \mathbb{Q}') = \Phi(d_1)$$
$$\lim_{N \rightarrow +\infty} \mathcal{B}(n \geq a | N, \mathbb{Q}) = \Phi(d_2)$$

18 The effect of a known dividend yield

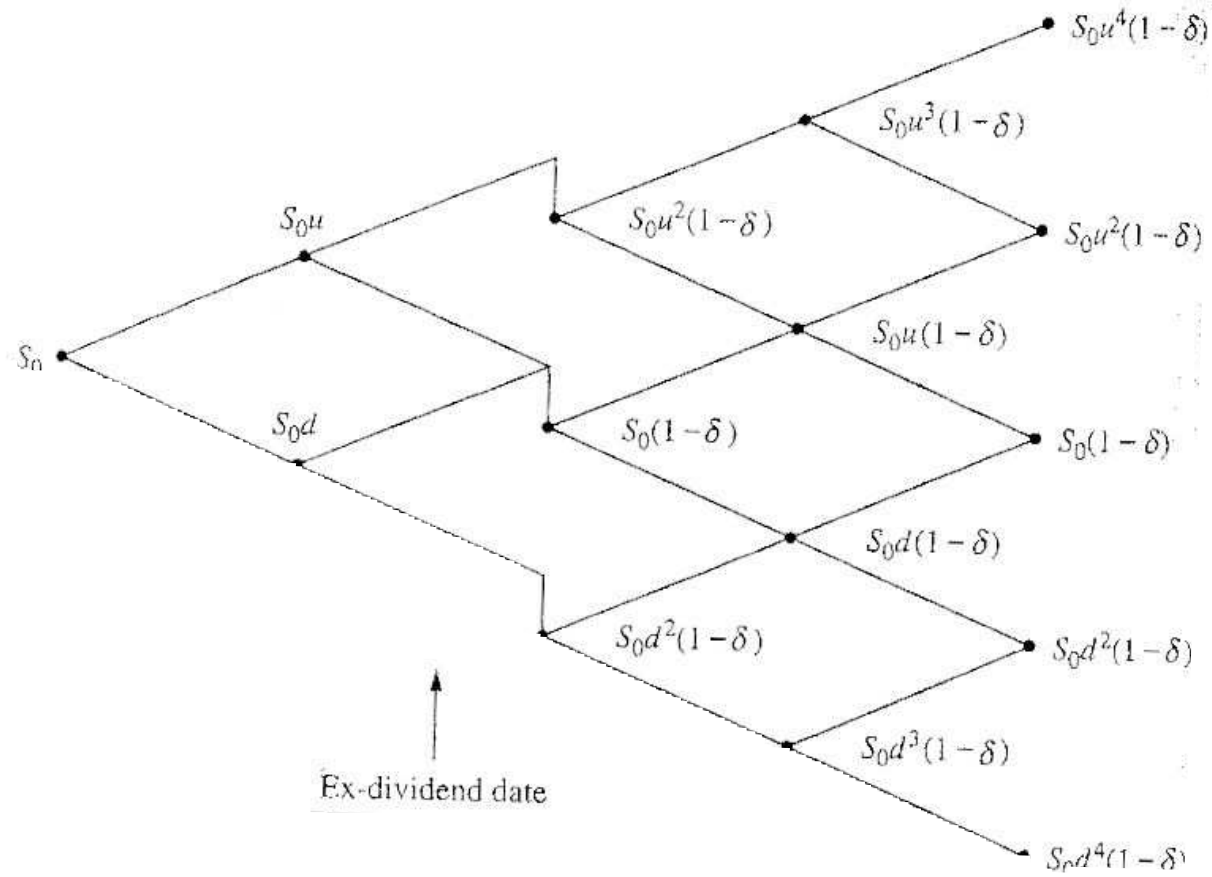
- In cases where the underlying asset pays a proportional dividend yield δ , the stock prices on tree nodes would be

$$S_{\underbrace{u \dots u}_j \underbrace{d \dots d}_{i-j}} = \begin{cases} S_0 u^j d^{i-j} & \text{for } i\Delta t \text{ before ex-div date} \\ S_0 (1 - \delta) u^j d^{i-j} & \text{for } i\Delta t \text{ after ex-div date} \end{cases}$$

with $i = 0, 1, \dots, N$ and $j = 0, 1, \dots, i$.

- The risk–neutral probability \mathbb{Q} in equation (7) is accordingly modified

$$\mathbb{Q} \equiv \frac{e^{(r-\delta)\Delta t} - d}{u - d} \quad (12)$$

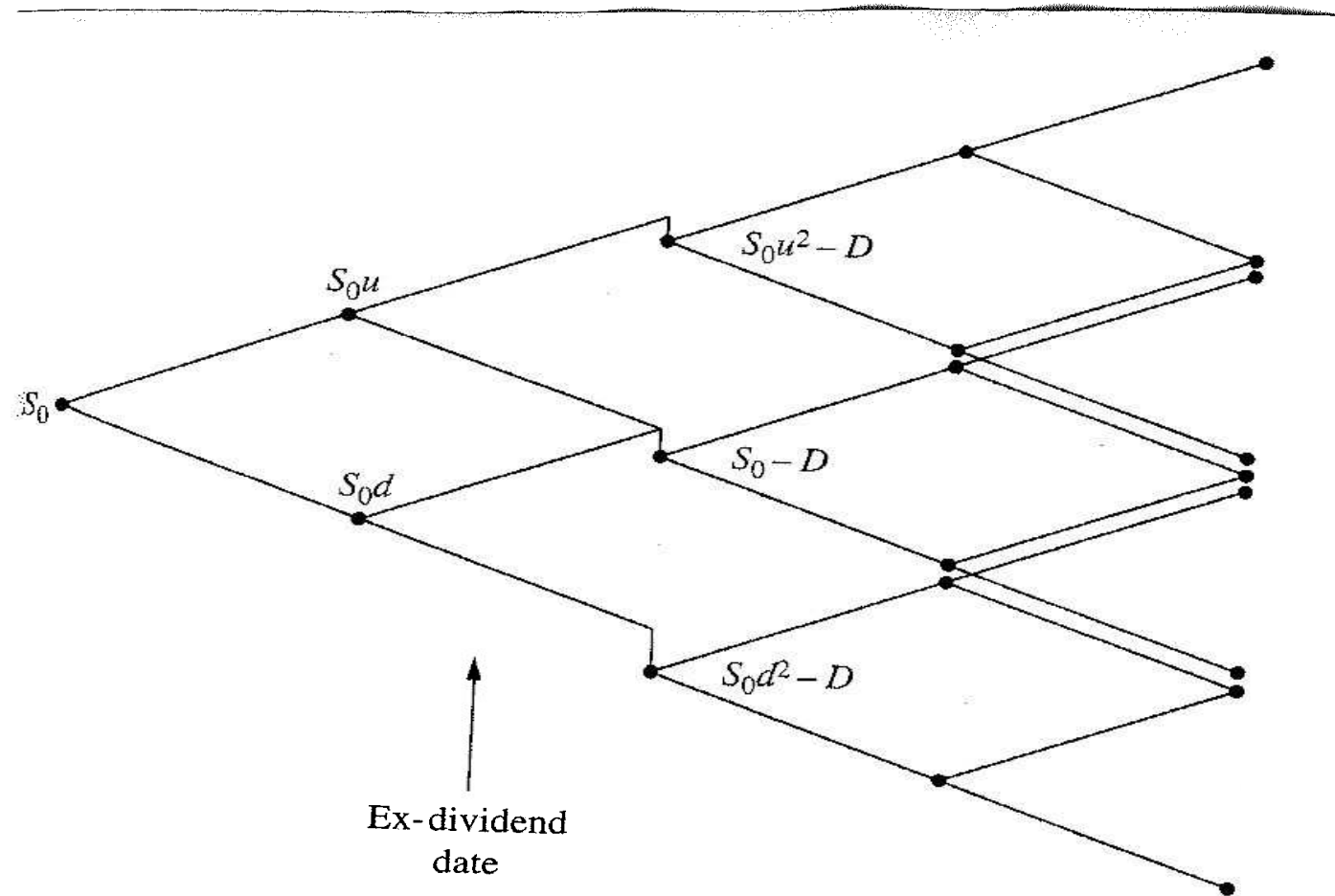


19 The effect of a known dollar dividend

- If the dividend is known as a dollar amount and not as a proportion of the stock price, then the resulting binomial tree is *non-recombining*, i.e. after the ex-div date the tree has $2i$ instead of $i + 1$ nodes. This increases the computational complexity considerably.
- In this case, the stock prices on tree nodes would be

$$S_{\underbrace{u \dots u}_j \underbrace{d \dots d}_{i-j}} = \begin{cases} S_0 u^j d^{i-j} & \text{for } i\Delta t \text{ before ex-div date} \\ (S_0 u^q d^{q-k} - D) u^{i-k} j^{i-j-k} & \text{for } i\Delta t \text{ after ex-div date} \end{cases}$$

with $i = 0, 1, \dots, N$, $j = 0, 1, \dots, i$, $q = 0, 1, \dots, k$ and k the lowest integer greater than the ex-div date.



20 Pricing American options in the binomial model

- An American–style option, unlike a European one, gives the holder the right (without the corresponding obligation) to buy (call, C) or sell (put, P) the underlying asset for a specific price, *at any point in time up until the specified maturity date*.
- In cases where the underlying asset pays no cash flows (e.g. dividends), we have shown (see lecture #3, p.11) that an American call will never be optimally early–exercised. However, the American put may be early exercised in such cases. How can we price an American put on a non–dividend paying asset?

- Firstly, we need to modify our backward–induction equation (9)

$$P_n = \max \left\{ \underbrace{K - S_n}_{\text{immediate}}, \underbrace{e^{-r\Delta t} \left[\mathbb{Q} P_{n+1}^{(u)} + (1 - \mathbb{Q}) P_{n+1}^{(d)} \right]}_{\text{put "alive"}} \right\} \quad (13)$$

- Again we would start from the option maturity $T = N \cdot \Delta t$, where

$$P_N = (K - \tilde{S}_N)^+$$

and work backwards to $t = 0$ using equation (13) as in the European case.

21 Two–period case

Let $N = 2$. At *maturity*:

$$\tilde{P}_2 = (K - \tilde{S}_2)^+ \quad (14)$$

where $S_2 \in \{S_2^{(uu)}, S_2^{(ud)}, S_2^{(dd)}\}$. At the *intermediate node*:

$$\tilde{P}_1 = \max \left\{ K - \tilde{S}_1, e^{-r\Delta t} \mathbb{E}^{\mathbb{Q}} [\tilde{P}_2 | \mathcal{F}_1] \right\} \quad (15)$$

where $S_1 \in \{S_1^{(u)}, S_1^{(d)}\}$. At the *onset*:

$$P_0 = e^{-r\Delta t} \mathbb{E}^{\mathbb{Q}} [\tilde{P}_1 | \mathcal{F}_0] \quad (16)$$

where $P_1 \in \{P_1^{(u)}, P_1^{(d)}\}$.

- Combine (14), (15) and (16) to get

$$P_0 = e^{-r\Delta t} \mathbb{E}^{\mathbb{Q}} \left[\underbrace{\max \left\{ (K - \tilde{S}_1)^+, e^{-r\Delta t} \mathbb{E}^{\mathbb{Q}} \left[\underbrace{(K - \tilde{S}_2)^+}_{\tilde{P}_2} \mid \mathcal{F}_1 \right] \right\}}_{\tilde{P}_1} \mid \mathcal{F}_0 \right]$$

- Thus to calculate the put price P_0 , we are interested only in the (conditional) expected present values of $(K - \tilde{S}_2)^+$ and $(K - \tilde{S}_1)^+$. To compute that we need to consider all possible “paths” of the asset price, i.e. couplets $\{S_1, S_2\}$.

- If the option is to be exercised, it will be:
 - Either at $t = \Delta t$ with payoff $(K - \tilde{S}_1)^+ > 0$
 - Or at $t = 2\Delta t$ with payoff $(K - \tilde{S}_2)^+ > 0$
and the random variable $\tau \in \{\Delta t, 2\Delta t\}$ is called a “stopping time”.

22 Numerical examples

- Consider the following set of parameters:[§]

The price of the underlying asset at $t = 0$

$$S_0 = 300$$

The strike price

$$K = 300$$

“up” factor

$$u = 2$$

“down” factor

$$d = \frac{1}{u} = \frac{1}{2}$$

time step

$$\Delta t = 1$$

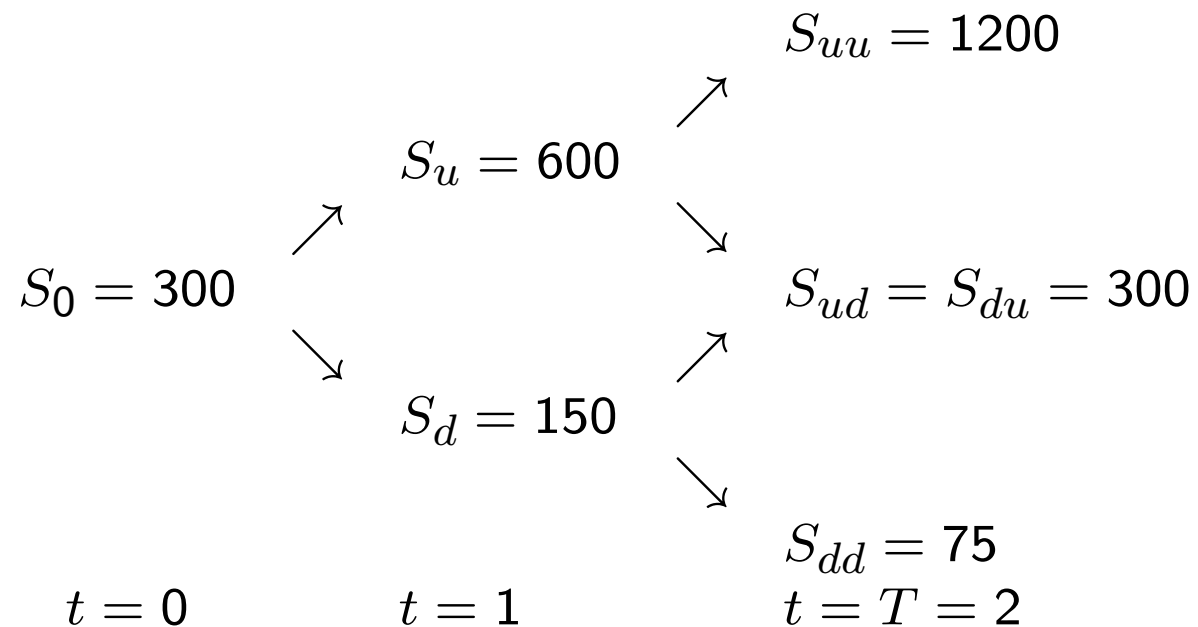
Discount factor

$$e^{-r\Delta t} = \frac{4}{5}$$

[§]Parameters values are rather unrealistic; however they should help demonstrate the principles.

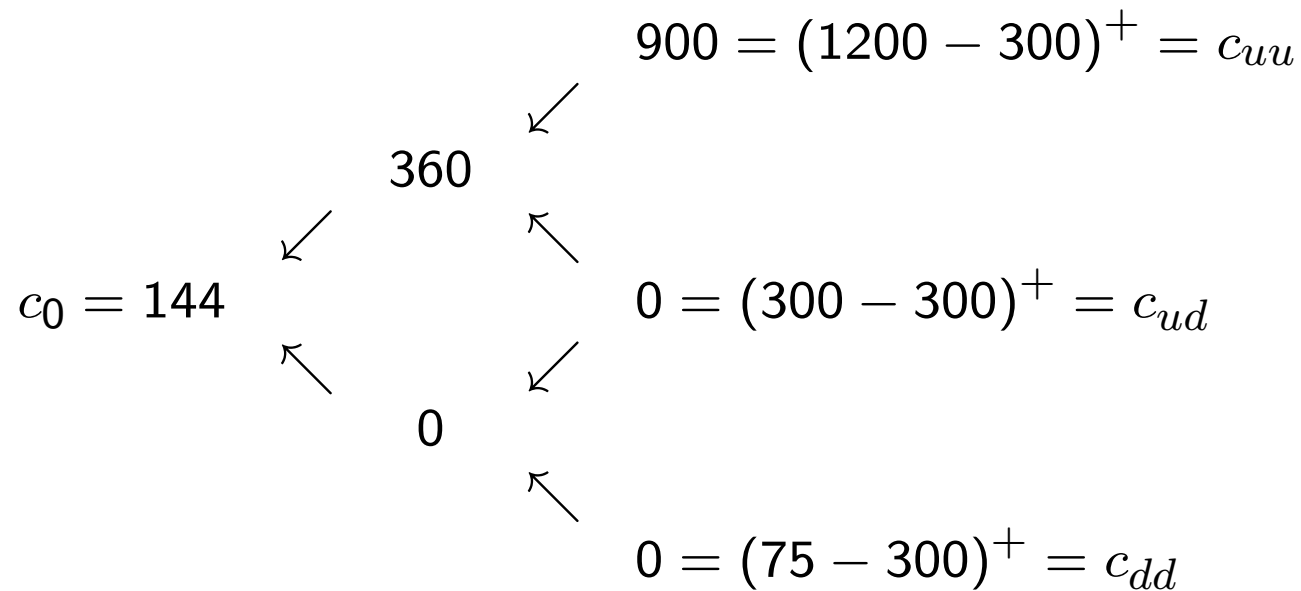
23 Preliminaries

- Build the underlying price “tree” and calculate the risk–neutral probabilities.



$$\mathbb{Q} \equiv \frac{e^{r\Delta t} - d}{u - d} = \frac{5/4 - 1/2}{2 - 1/2} = \frac{1}{2} \quad 1 - \mathbb{Q} = 1 - \frac{1}{2} = \frac{1}{2}$$

24 European call



$$c_u = e^{-r\Delta t} [Qc_{uu} + (1 - Q)c_{ud}] = 360$$

$$c_0 = e^{-r\Delta t} [Qc_u + (1 - Q)c_d] = 144$$

25 American call

$$C_0 = 144$$

$$360 = \max \left\{ \underbrace{60}_{S_u - K}, 360 \right\}$$

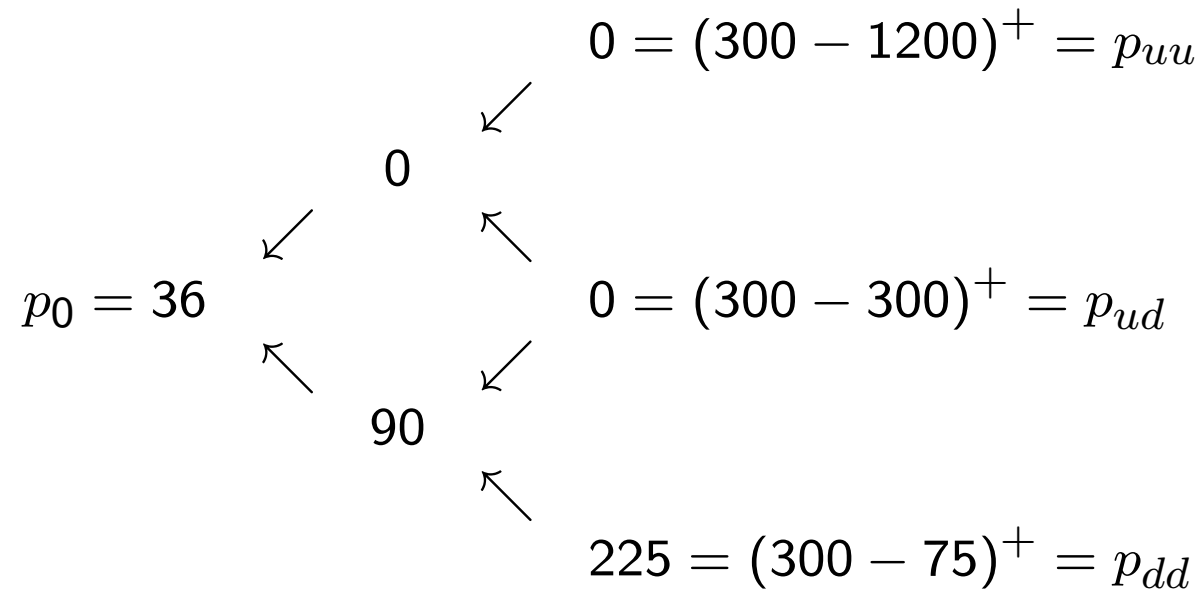
$$0 = \max \left\{ \underbrace{-150}_{S_u - K}, 0 \right\}$$

$$900 = (1200 - 300)^+ = C_{uu}$$

$$0 = (300 - 300)^+ = C_{ud}$$

$$0 = (75 - 300)^+ = C_{dd}$$

26 European put



$$p_d = e^{-r\Delta t} [\mathbb{Q}p_{uu} + (1 - \mathbb{Q})p_{ud}] = 90$$

$$p_0 = e^{-r\Delta t} [\mathbb{Q}p_u + (1 - \mathbb{Q})p_d] = 36$$

27 American put

$$\begin{array}{r}
 P_0 = 60 \\
 \swarrow \quad \nwarrow \\
 0 = \max \left\{ \underbrace{-300}_{K-S_u}, 0 \right\} \\
 \swarrow \quad \nwarrow \\
 150 = \max \left\{ \underbrace{150}_{K-S_u}, 90 \right\} \\
 \swarrow \quad \nwarrow \\
 0 = (300 - 1200)^+ = P_{uu} \\
 \swarrow \quad \nwarrow \\
 0 = (300 - 300)^+ = P_{ud} \\
 \swarrow \quad \nwarrow \\
 225 = (75 - 300)^+ = P_{dd}
 \end{array}$$

- Note that

$$\underbrace{P_0}_{\text{American}} = 60 > 36 = \underbrace{p_0}_{\text{European}}$$

$$\underbrace{C_0}_{\text{American}} = 144 = 144 = \underbrace{c_0}_{\text{European}}$$

28 Advantages and disadvantages

- Advantages of the binomial model:
 - Can easily handle situations where there are early exercise opportunities.
 - Can be used when the payoff of the contract depends on the actual path followed by the underlying variable S , as well as when it depends only on the final value of S .
 - Demonstrates clearly and in an easy framework the concepts of no-arbitrage pricing and risk neutral valuation.
- Disadvantages of the binomial model:
 - It is impossible to extend the model in situations where the option from the derivative depends on more than two underlying market variables.
 - Not all stochastic processes for the underlying asset S can be accommodated.

29 Reading

- Hull [5], Chapters 10, 18 (pp. 392–414), 19–20 (advanced reading)
- Baxter and Rennie [1], Chapter 2
- Jarrow and Turnbull [6], Chapters 5, 17, 19, 20
- Stulz [8], chapter 12
- Cox, Ross and Rubinstein [4]

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