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LEDa

Chapter 2

#### Introduction Motivation

- A useful and very popular technique for pricing an option involves constructing a binomial tree.
	- This is a diagram representing different possible paths that might be followed by the stock price over the life of an option.
	- The underlying assumption is that the stock price follows a random walk



#### Basic notions on Probability (Part 1)

- A useful and very popular technique for pricing an option involves constructing a binomial tree.
- Let  $(\Omega, \mathcal{F}, \mathbb{P})$  denotes a *probability space*. That is, a triple of:
	- $\triangleright$   $\Omega$  a sample space which is the universe of possible outcomes:
	- $\triangleright$  *F* a set of events, where an event is a subset of  $\Omega$ ;
	- $\triangleright$  P a probability function from F to [0, 1], which measures the likeliness that an event will occur

#### Basic notions on Probability (Part 1)

- $\bullet$  Observe that for the probability  $\mathbb P$  to be well-defined, the set of events  $F$  has to satisfy some properties:
	- $\triangleright$  *F* is non-empty:
	- $\triangleright$  F is closed under complementation: If A is in F, then so is its complement,  $\Omega \backslash A$ ; and
	- $\triangleright$  F is closed under countable unions: If  $A_1$ ,  $A_2$ ,  $A_3$ , ... are in F, then so is  $A = A_1 \cup A_2 \cup A_3 \cup ...$
- We say that such F is a  $\sigma$ -algebra (or  $\sigma$ -field).
	- In general, we will take  $\mathcal F$  as the smallest  $\sigma$ -algebra generated by the experiment.



## Basic notions on Probability (Part 1)



Basic notions on Probability (Part 1)

#### Example (A)

We consider the experiment that consists in rolling a dice and then checking whether the number 6 is the outcome.

So,  $\Omega = \{1, 2, 3, 4, 5, 6\}$  and  $\mathcal{F}_A = \{\emptyset, \{6\}, \{1, 2, 3, 4, 5\}, \Omega\}.$ 

#### Example (B)

We consider the experiment that consists in rolling a dice and then checking whether the outcome is even.

So,  $\Omega = \{1, 2, 3, 4, 5, 6\}$  and  $\mathcal{F}_B = \{\emptyset, \{2, 4, 6\}, \{1, 3, 5\}, \Omega\}.$ 

## Basic notions on Probability (Part 1)

#### Example (C)

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We consider the experiment that consists in rolling a dice and then checking the outcome.

So,  $\Omega = \{1, 2, 3, 4, 5, 6\}$  and  $\mathcal{F}_C = 2^{\Omega}$ , where  $2^{\Omega}$  denotes the power set of the sample space.

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I.e.,  $\mathcal{F}_C$  has  $2^6 = 64$  elements. E.g., one of this element is  $\{2, 5\}$ , which consists in checking whether the outcome is 2 or 5.

## **Binomial tree (Part 1)**

- We consider a market with only two periods:  $t = 0$  and  $t = 1$ .
- There are two assets.
	- A riskless asset who values 1 at date  $t = 0$  and  $R = (1 + r)$  at date  $t = 1$ . r denotes the risk-free interest rate that we could obtain with a zero coupon.
	- A risky asset S who values  $S_0$  at date  $t = 0$  and can take two different values at date  $t = 1$ :  $S_1 \in \{S_1^u, S_1^d\}$  with  $S_1^u = uS_0$ ,  $S_1^d = dS_0$ , and  $d < u$ .

#### Example (D)





## **Binomial tree (Part 1)**

- Let  $(\Omega, \mathcal{F}, \mathbb{P})$  denotes the probability space corresponding to this market situation. We have:
	- $\bullet \ \Omega = {\omega_{\mu}, \omega_{\sigma}};$
	- $\triangleright$   $\mathcal{F}_0 = {\varnothing, \Omega}$ ;  $\mathcal{F}_1 = {\varnothing, {\{\omega_{\mu}\}, {\{\omega_{\mu}\}, \Omega\}}}$ ; and
	- Figure 1 =  $\mathbb{P}$  such that  $\mathbb{P}(\omega_{\mu}) = p$  and  $\mathbb{P}(\omega_{\mu}) = 1 p$ , with  $p \in (0, 1)$ .
- Observe that  $\mathcal{F}_0 \subset \mathcal{F}_1$ : we acquire information through time.



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#### Chapter 2: Binomial tree with one period Outline

#### Introduction

- Basic notions on Probability (Part 1)
- **Binomial tree (Part 1)**
- Basic notions on Probability (Part 2)
- **Binomial tree (Part 2)** • Simple portfolio strategies
- Basic notions on Probability (Part 3)
- **Binomial tree (Part 3)** • Evaluating and hedging derivative

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## Basic notions on Probability (Part 2)

#### Definition

Let  $(\Omega, \mathcal{F}, \mathbb{P})$  be a probability space. A (real-valued) random variable is a (real) function  $X : \Omega \longmapsto \mathbb{R}$  such that  $\{\omega \in \Omega | X(\omega) \le x\} \in \mathcal{F}$  for every  $x \in \mathbb{R}$ .

• Said differently, a random variable is a function that assigns a numerical value to each state of the world,  $X : \Omega \longmapsto \mathbb{R}$ , such that the values taken by  $X$  are known to someone who has access to the information  $\mathcal F$ .

#### Example (A')

The window will be opened if and only if the maximal number of the dice is realized.

By associating the number 1 to the action of opening the window and zero otherwise, we have:

$$
\mathsf{X}_{\mathsf{A}} = \left\{ \begin{array}{l} \mathsf{1} \text{ if } \omega = \{\mathsf{6}\} \\ 0 \text{ otherwise.} \end{array} \right.
$$

Observe that we can use  $\mathcal{F}_A$  (or  $\mathcal{F}_C$ ) but not  $\mathcal{F}_B$ .

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## Basic notions on Probability (Part 2)

#### Example (B')

You earn 100 $\in$  if the realization of the dice is even and you lose 50 $\in$ otherwise.

$$
\mathsf{X}_\mathsf{B} = \left\{\begin{array}{c} +100 \text{ if } \omega \in \{2,4,6\} \\ -50 \text{ otherwise.} \end{array}\right.
$$

Observe that we can use  $\mathcal{F}_B$  (or  $\mathcal{F}_C$ ) but not  $\mathcal{F}_A$ .

#### Example (C')

You earn  $15 \in$  if the realization of the dice is 5 and zero otherwise.

$$
\zeta_C = \left\{ \begin{array}{c} +15 \text{ if } \omega = \{5\} \\ 0 \text{ otherwise.} \end{array} \right.
$$

Observe that we can use  $\mathcal{F}_C$  but neither  $\mathcal{F}_A$  nor  $\mathcal{F}_B$ .

Let F denotes a  $\sigma$ -algebra associated with  $\Omega$ .

A (real) function  $X : \Omega \rightarrow \mathbb{R}$  is  $\mathcal{F}-$ **measurable** if, for any two numbers  $a, b \in \mathbb{R}$ , all the states of the world  $\omega \in \Omega$  for which X takes value between a and b forms a set that is an event (an element of  $\mathcal{F}$ ).

Formally,  $\forall a, b \in \mathbb{R}$ ,  $a < b$ , we have  $\{\omega \in \Omega | a < X(\omega) < b\} \in \mathcal{F}$ .

- So, a random variable is  $F$  measurable if and only if it is known with the information given by  $\mathcal F$ .
	- I.e., for any two numbers, we are able to answer the question on whether the realization of the random variable belongs to the interval formed by these two numbers.
	- ► Roughly speaking, we are able to say what actually happened.

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## Basic notions on Probability (Part 2)

#### Example (A")

 $X_A$  is  $\mathcal{F}_A$ -measurable and  $\mathcal{F}_C$ -measurable but is not  $\mathcal{F}_B$ -measurable.

#### Example (B")

 $X_B$  is  $\mathcal{F}_B$ -measurable and  $\mathcal{F}_C$ -measurable but is not  $\mathcal{F}_A$ -measurable.

#### Example (C")

 $X<sub>C</sub>$  is  $\mathcal{F}<sub>C</sub>$ -measurable but neither  $\mathcal{F}<sub>A</sub>$ -measurable nor  $\mathcal{F}_B$ -measurable.

## Basic notions on Probability (Part 2)

- A more general definition is that a function  $X: G \rightarrow H$  is measurable if the preimage under  $X$  of every element in the  $\sigma$  – algebra associated with H is in the  $\sigma$  – algebra associated with G.
	- Formally, if G (resp. H) is the  $\sigma$ -algebra associated to G (resp. H), then  $X^{-1}(v) := \{q \in G | X(q) = v \} \in \mathcal{G}, \forall v \in \mathcal{H}.$
	- The idea is that a measurable function pulls back measurable sets.
- The notion of measurability depends on the  $\sigma$ -algebras that are used.
	- In our definition, as the  $\sigma$ -algebra associated with R we took the Borel  $\sigma$ -algebra on the reals, i.e., the smallest  $\sigma$ -algebra on R which contains all the intervals

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#### Chapter 2: Binomial tree with one period **Outline**

#### Introduction

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- Basic notions on Probability (Part 1)
- **Binomial tree (Part 1)**
- Basic notions on Probability (Part 2)
- **Binomial tree (Part 2)** • Simple portfolio strategies

**Basic notions on Probability (Part 3)** 

#### **Binomial tree (Part 3)**

Evaluating and hedging derivative

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# • The risky asset S (who values  $S_0$  at date  $t = 0$  and can take two<br>different values at date  $t = 1$ :  $S_1 \in \{S_1^u, S_1^d\}$  is a random variable<br>that is  $\mathcal{F}_1$ —measurable, but is not  $\mathcal{F}_0$ —measurable.<br>
• That is, **•** That is, the information known at date 0 is not sufficient to say what<br>
is the realization of S.<br>
• Instead, we have to wait until date 1.<br>
• Observe that  $\mathcal{F}_1$  is the smallest  $\sigma$ -algebra that makes S<br>
measurabl Stock price = \$20<br>
Option Price = \$1<br>
Option Price = \$1<br>
Option Price = \$1<br>
Stock Price = \$1<br>
Option Price • If  $S_1$  is the underlying asset, then any derivative can be written as a  $\mathcal{F}_1$ -measurable function of S<sub>1</sub>. Example A call with underlying x and strike  $K$  is a derivative that takes the form

 $\phi: x \longmapsto (x - K)^{+}.$ 

$$
X_0^{x,\Delta} = \Delta S_0 + (x - \Delta S_0) \mathbf{1} = x. \tag{1}
$$

$$
X_1^{x,\Delta} = \Delta S_1 + (x - \Delta S_0) R = xR + \Delta (S_1 - S_0 R).
$$

## Simple portfolio strategies

#### Theorem (2.1)

In our market, every derivative is replicable by using a simple portfolio strategy  $(x, \Delta)$ .





Simple portfolio strategies

• So, under NAO, the price of a derivative in period  $t = 0$  is given by

$$
C_0 = X_0^{x,\Delta} = x
$$
  
= 
$$
\frac{1}{R} \left( \frac{R-d}{u-d} C_1^u + \frac{u-R}{u-d} C_1^d \right)
$$

which is a weighted sum of its future values  $C_1^u$  and  $C_1^d$ .

#### Example (D")

Assume the 3 months risk-free rate is 3.05%. We then obtain

$$
C_0=\frac{1}{1.0305}\left(\frac{1.0305-0.9}{1.1-0.9}\right)\simeq 0.633
$$

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- A market where every asset is replicable with a simple portfolio strategy is said to be complete.
- Now let us study how the initial value of a simple portfolio strategy,  $X_0^{x,\Delta}$ , depends on its future value,  $X_1^{x,\Delta}$ .

A simple arbitrage is a simple portfolio strategy that gives to a portfolio no value at time  $t = 0$  and a value at time  $t = 1$  which is strictly positive with positive probability and is never negative.

Formally, it is a pair  $(x = 0, \Delta)$  with  $\Delta \in \mathbb{R}$  such that

 $X_1^{0,\Delta} \ge 0$  and  $\mathbb{P}(X_1^{0,\Delta} > 0) > 0$ .

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**Binomial tree (Part 2)** Simple portfolio strategies

#### **Definition**

We say that there is no simple arbitrage opportunity (NAO') if

$$
\forall \Delta \in \mathbb{R}, \ \{X_1^{0,\Delta} \geq 0 \implies X_1^{0,\Delta} = 0 \ \mathbb{P}-a.s.\}
$$

#### Proposition (2.2)

If NAO' then  $d < R < u$ .





Simple portfolio strategies



$$
\tilde{X}^{\mathsf{x}, \Delta}_t := \frac{X^{\mathsf{x}, \Delta}_t}{R^t}.
$$

 $\tilde{X}_{0}^{X,\Delta}=x$ 

 In term of current values, the portfolio self-financing condition then

$$
\tilde{X}^{x,\Delta}_1-\tilde{X}^{x,\Delta}_0=\Delta\left(\tilde{S}_1-\tilde{S}_0\right).
$$

collection of random variables on  $\Omega$ , indexed by a totally ordered set  $\mathcal{T}$ <br>(e.g., referring to time).<br>Formally, a stochastic process  $X$  is a collection  $(X_t)_{t \in \mathcal{T}}$  where each  $X_t$ <br>is a random variable on  $\Omega$ .<br>• time).<br>
aastic process X is a collection  $(X_t)_{t \in T}$  where each  $X_t$ <br>
able on  $\Omega$ .<br>  $2, ..., n$  the stochastic process is discrete. We will<br>  $X_k)_{1 \leq k \leq n}$ .<br>
Aralitage&Priding<br> **a** on Probability (Part 3)<br>
iminary)<br>
a stoch Chapter 2: Binomial tree with one period Outline **Introduction** Basic notions on Probability (Part 1) **Binomial tree (Part 1)** Basic notions on Probability (Part 2)  $n$  we have **Binomial tree (Part 2)**  $\mathbb{E} \left[ |X_n| \right] < +\infty$ • Simple portfolio strategies and Basic notions on Probability (Part 3)  $E[X_{n+1}|X_1,...,X_n]=X_n$ **Binomial tree (Part 3)** • Evaluating and hedging derivative

#### Definition

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- **CERT AS THE FIRST WARE THE FIRST WAT AS THE FIRST WIND SEVERALLY (Part 3)**<br>
Samally, martingale referred to a class of betting strategies that<br>
simplest of these strategies was designed for a game in which<br>
simplest of th **Example in the gambler of the gambler of the gambler with and available time joint and available time jointly approach infinity, the stategies were stategies were stategies was designed for a game in which<br>
The idea is th CONDITY (Part 3)**<br>
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simplest of these strategies was designed for a game in which<br>
simplest of these strategies was designed for a game in which<br>
simplest
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- sic notions on Probability (Part 3)<br>The idea is that an "equivalent martingale" measure is a probability<br>measure under which the current value of all financial assets at time<br>t is equal to the expected future payoff of th
	-

stake if a coin comes up heads and loses it if measure under which the current value of all<br>
is.<br>
It is equal to the expected future payoff of the<br>
recover all previous losses plus win a profit equal<br>
recover all previous

the first win would recover all previous losses plus win a profit equal<br>
to the original stake.<br>
As the gambler's wealth and available time jointly approach infinity,<br>
his probability of eventually flipping heads approach

 $\widetilde{X}^{\scriptscriptstyle{X,\Delta}}_0 = \mathbb{E}^{\mathbb{Q}} \left[ \widetilde{X}^{\scriptscriptstyle{X,\Delta}}_1 \right]$ 

or equivalently

$$
X_0^{x,\Delta} = \frac{1}{R} \mathbb{E}^{\mathbb{Q}} \left[ X_1^{x,\Delta} \right].
$$

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- **Binomial tree (Part 2)** • Simple portfolio strategies

Basic notions on Probability (Part 3)

#### **Binomial tree (Part 3)**

• Evaluating and hedging derivative

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## **Binomial tree (Part 3)**

#### Proposition (2.3)

If  $d < R < u$  then there is an equivalent martingale measure Q.





## **Binomial tree (Part 3)**



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 $\Box$ 

• Following the two previous propositions, we have

 $NAO' \Longrightarrow d < R < u \Longrightarrow$  there is an equivalent martingale measure  $\Longrightarrow NAO'$ .

- Hence we obtain
	- $NAO' \iff d < R < u \iff$  there is an equivalent martingale measure.

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**Binomial tree (Part 3)** Evaluating and hedging derivative

- Observe that the equivalent martingale measure does not depend on the probabilities p (and  $1-p$ ) of the state  $\omega_{\mu}$  (and  $\omega_{d}$ ).
	- $\triangleright$  So, the price of an option is independent from the probability behind the evolution of the underlying asset.
		- $\star$  This is partly due to the fact that the replicating portfolio contains the underlying asset.
	- $\triangleright$  To determine the price of the derivative we then just need to know r.  $u$ , and  $d$ .

#### Question

How to determine  $u$  and  $d$ ?

• We shall see how this is correlated with the volatility of the asset.



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**Binomial tree (Part 3)** Evaluating and hedging derivative

• In the replicating portfolio, the quantity of the risky asset is given by

$$
\Delta = \frac{C_1^u - C_1^d}{(u-d) S_0} = \frac{\phi(S_1^u) - \phi(S_1^d)}{(u-d) S_0}.
$$

 $\triangleright$  This quantity measures how the price of the option varies with the underlying asset price variation.



## **Binomial tree (Part 3)**

Evaluating and hedging derivative

#### Proposition (2.5)

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Assume NAO. The price of a derivative at time  $t = 0$  is given by

$$
C_0=\frac{\mathbb{E}^{\mathbb{Q}}\left[C_{1}\right]}{1+r}=\frac{1}{R}\left(\mathbb{Q}\left(\omega_{u}\right)C_{1}^{u}+\mathbb{Q}\left(\omega_{d}\right)C_{1}^{d}\right)=\frac{1}{R}\left(qC_{1}^{u}+\left(1-q\right)C_{1}^{d}\right)
$$

#### Proof.

Exercise. (Hint: Straightforwardly obtained from the previous section)

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 $\Box$ 

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#### **Binomial tree (Part 3)** Evaluating and hedging derivative

#### Example  $(D<sup>(4)</sup>)$

The hedging strategy consists then:

- in buying 0.25 unit of the risky asset (the cost is  $\Delta S_0 = 0.25 \times 20 = 5$ ; and

 $\frac{0.0305-0.9}{0.0305-0.9} = 5 \sim -4.3668$  in  $.0305$  1.1–0.9  $\sim$  –  $.1-0.9$   $.0000$ non-risky asset. Doing so, we indeed obtain

$$
-4.3668\times1.0305+0.25\times22=1=C_1^u
$$

and

$$
-4.3668\times1.0305+0.25\times18\simeq0={C_1^d}.
$$

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## **Binomial tree (Part 3)** Evaluating and hedging derivative

- An alternative way to determine  $\Delta$  is to consider a portfolio consisting of a long position in  $\Delta$  shares of the risky asset and a short position in one call option, and then to calculate the value of  $\Delta$ that makes this portfolio riskless.
	- If there is an up movement in the stock price, the value of the portfolio at the end of the life of the option is

$$
S_0 u\Delta-C_1^u
$$

If there is a down movement in the stock price, the value becomes

$$
\mathcal{S}_0d\Delta-C_1^d
$$

▶ The two are equal (i.e., 
$$
S_0 u\Delta - C_1^u = S_0 d\Delta - C_1^d
$$
) when

$$
\Delta = \frac{C_1^u - C_1^d}{(u - d) S_0}.
$$

\n- at time 1 is 
$$
S_0 u \Delta - C_1^u
$$
 (=  $S_0 d \Delta - C_1^d$ );
\n- today is  $\frac{S_0 u \Delta - C_1^u}{1 + r}$ ;
\n

- 
- Hence

$$
C_0=S_0\Delta-\frac{S_0u\Delta-C_1^u}{1+r}
$$

**Substituting for** 
$$
\Delta = \frac{C_1^u - C_1^d}{S_0 u - S_0 d}
$$
 we obtain

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\nUsing derivative

\nportfolio:

\n
$$
u\Delta - C_{1}^{u} (= S_{0}d\Delta - C_{1}^{d});
$$
\n
$$
C_{1}^{u};
$$
\non for the portfolio value today is

\n
$$
S_{0}\Delta - C_{0}
$$
\n
$$
C_{0} = S_{0}\Delta - \frac{S_{0}u\Delta - C_{1}^{u}}{1+r}
$$
\n
$$
= \frac{C_{1}^{u} - C_{1}^{d}}{S_{0}u - S_{0}d}
$$
\nwe obtain

\n
$$
C_{0} = \frac{qC_{1}^{u} + (1-q)C_{1}^{d}}{1+r}
$$
\n, which confirms Proposition 2.5.

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\nPart 3)

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## Evaluating and hedging derivative

#### Proposition (2.6)

If every asset is replicable with a simple portfolio strategy (complete market) then the equivalent martingale measure is unique.

# Proof.  $\Box$



- Binomial trees illustrate the general result that to value a derivative we can assume that:
	- $\triangleright$  The expected return on a stock (or any other investment) is the risk-free rate.
	- $\triangleright$  The discount rate used for the expected payoff on an option (or any other instrument) is the risk-free rate.
- This is known as using risk-neutral valuation.



## Evaluating and hedging derivative

 $\bullet$  q is the probability that gives a return on the stock equal to the risk-free rate:

$$
S_0(1+r) = S_1^u q + S_1^d(1-q).
$$

• The value of the option is

$$
C_0 = \frac{C_1^u q + C_1^d (1 - q)}{1 + r}
$$

**Binomial tree (Part 3)** 

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**Evaluating and hedging derivative** 

• It is natural to interpret q and  $1-q$  as probabilities of up and down movements.

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- The value of a derivative is then its expected payoff in a risk-neutral world discounted at the risk-free rate.
- When the probability of an up and down movements are  $q$  and  $1-q$  the expected stock price at time 1 is  $S_0(1+r)$ .
- This shows that the stock price earns the risk-free rate.

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#### Example  $(D^{(5)})$

We have

 $20(1.0305) = 22q + 18(1 - q).$ 

so that  $q = 0.6525$ . And

 $C_0 = \frac{1 \times 0.6525 + 0(1 - 0.6525)}{1.0305} \simeq 0.6332.$ 



#### **Binomial tree (Part 3)** Evaluating and hedging derivative

#### Question

What is the Call price of a Call with  $S_0 = 100$ ,  $K = 100$ ,  $r = 0.05$ ,  $d = 0.9$  and  $u = 1.1$ ?

Give a hedging strategy and depict a tree that illustrates the replication.

#### Solution

**Binomial tree (Part 3)** Evaluating and hedging derivative



**Binomial tree (Part 3)** Evaluating and hedging derivative



## Binomial tree (Part 3)

Evaluating and hedging derivative

#### Question

What about a Put with the same characteristics?



**Binomial tree (Part 3)** Evaluating and hedging derivative

#### Question

Does the Call-Put parity holds?



**Binomial tree (Part 3)** Evaluating and hedging derivative

