

# Lecture 6

## Option Pricing in Binomial Model

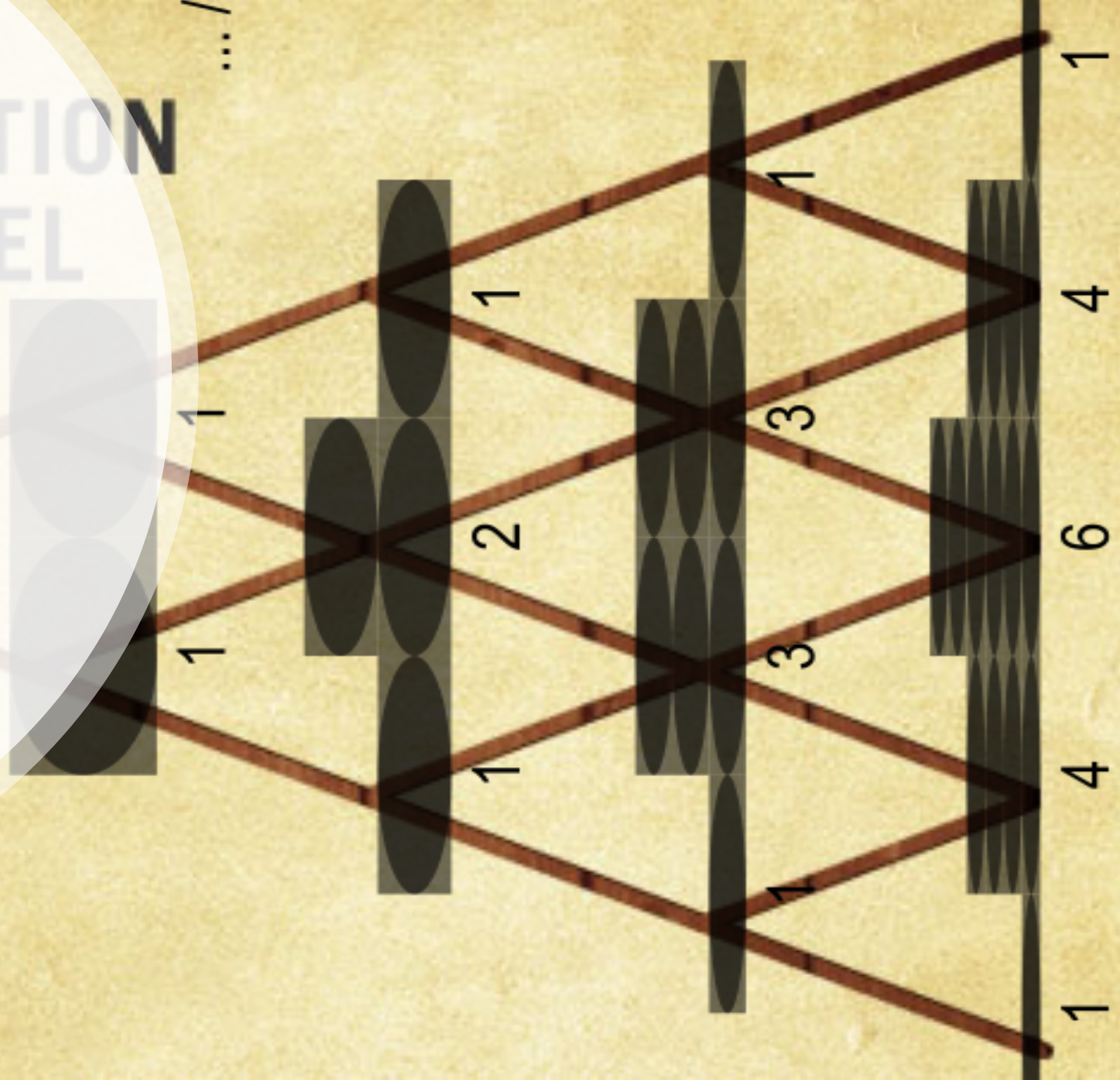
Financial instruments and pricing

Fall 2019

# Option pricing in Binomial Model

## ❖ Stochastic processes

- ❖ One-step Binomial Model
- ❖ Multi-step Binomial Model
- ❖ “Analytic” formula for European options
- ❖ Risk-neutral / martingale pricing



# Stochastic processes (short introduction)

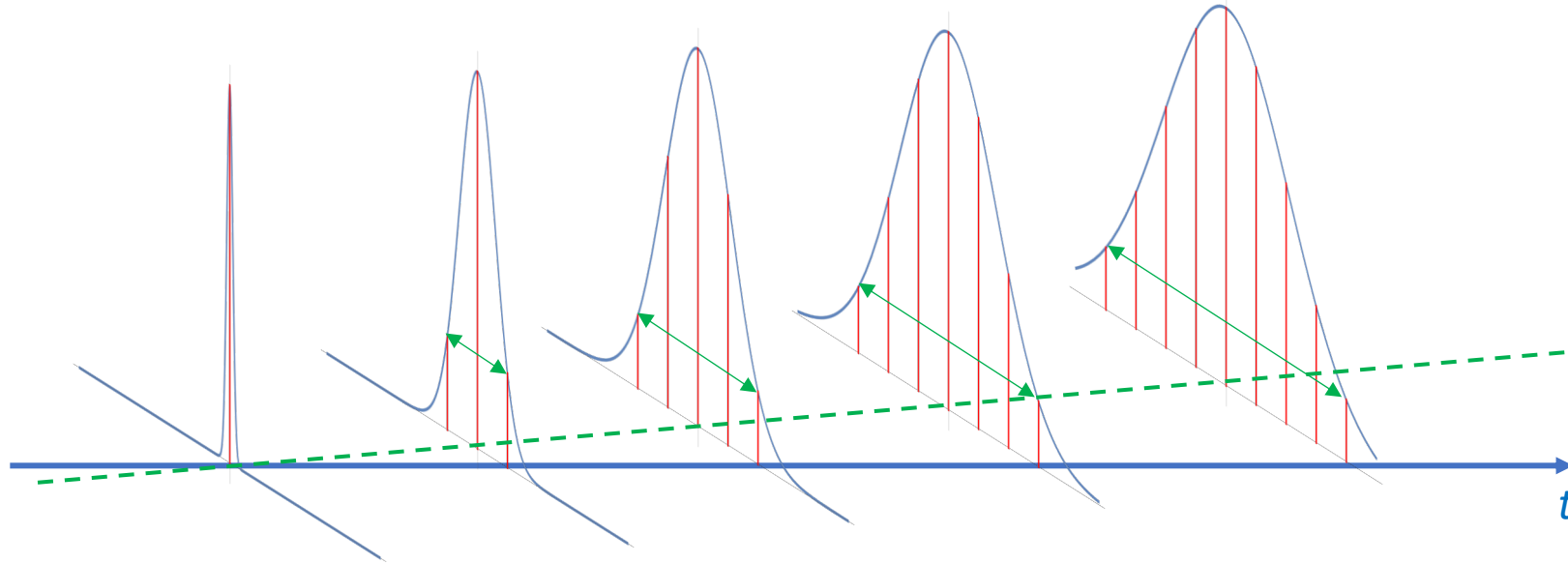
- ❖ As already discussed in **Lecture 5**, in order to price **RISK ASYMMETRIC** derivatives (**OPTIONS**) one has to assume some **STOCHASTIC PROCESS** driving the underlying asset (e.g. share) **PRICE EVOLUTION**
- ❖ If the **MARKET IS DYNAMICALLY COMPLETE**, i.e. if one can replicate the derivative instrument using some **DYNAMICALLY ADJUSTED PORTFOLIO**, then (under some additional conditions of risk-neutral / martingale price evolution) one can use "standard" pricing methods of eqn. (4)

$$\sum_t < PV(CF(t)) > = 0 \quad (4)$$

- ❖ Let's start by explaining what actually **STOCHASTIC PROCESSES** are ?

# Stochastic processes (short introduction)

❖ A **stochastic process** is a set (family) of **random variables indexed by time**:  $\{S(t)\}_t$

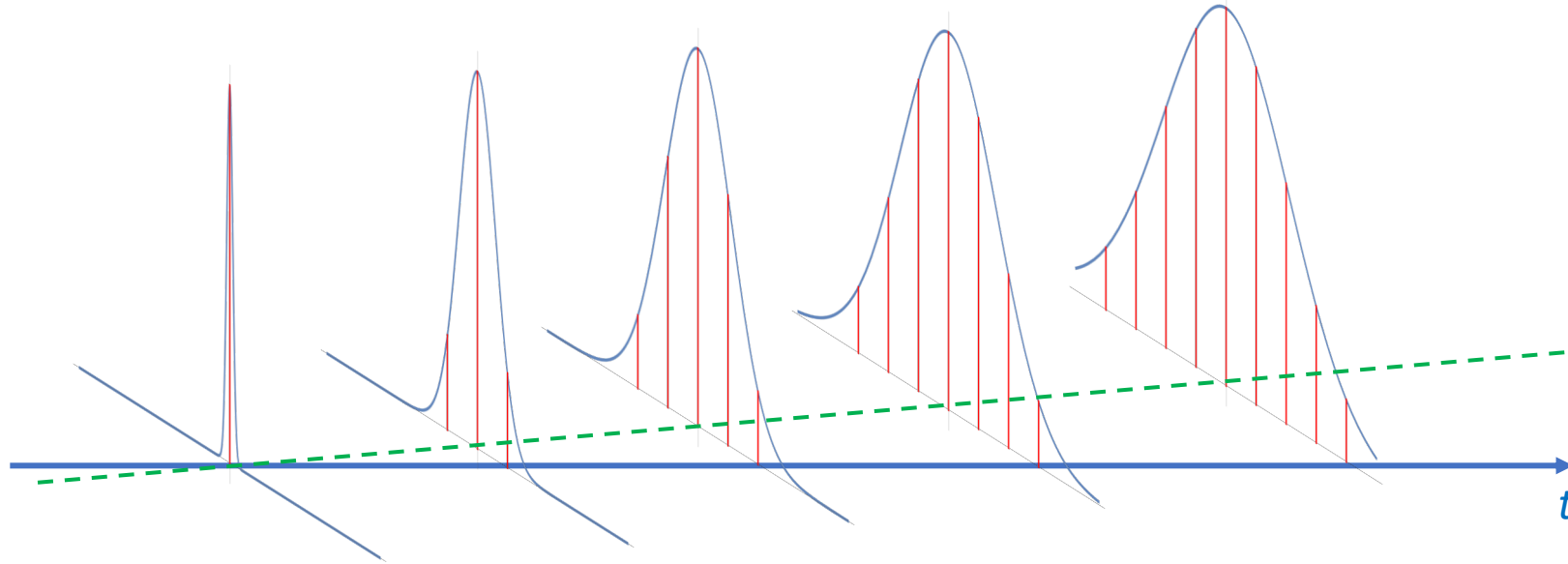


❖ Usually the random variables  $S(t)$  observed in various moments in time are assumed to have the **SAME TYPE of PROBABILITY DISTRIBUTION (FUNCTION)**, but **PARAMETERS** (e.g. **mean**, **variance**, ...) **may CHANGE / EVOLVE IN TIME\***:  $S(t) \sim \text{PDF}(\mu(t), \dots)$

\* If the random variables are identically distributed  $\forall t$ , then the process is called “**stationary**”. Parameters change for „**non-stationary**” processes.

# Stochastic processes (short introduction)

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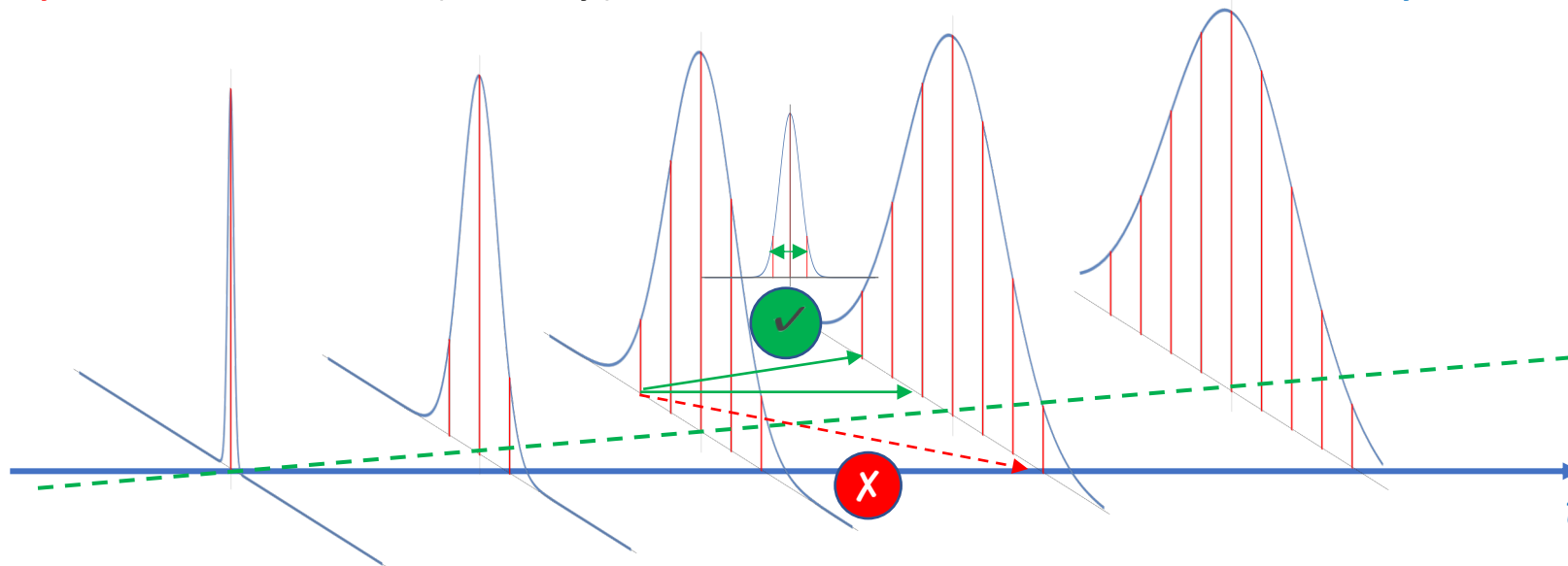


❖ Both the **random variables**  $S(t)$  and **time index**  $t$  can be either **discrete** or **continuous**, so one can have:

- **discrete time** - **discrete „space“** (e.g. **Binomial**) process
- **continuous time** - **continuous „space“** (e.g. **Wiener** ⇐ will be discussed in **Lecture 7**) process,
- **continuous time** - **discrete „space“** (e.g. **Poisson**) process
- **discrete time** - **continuous „space“** (e.g. **ARMA**, **GARCH**) process

# Stochastic processes (short introduction)

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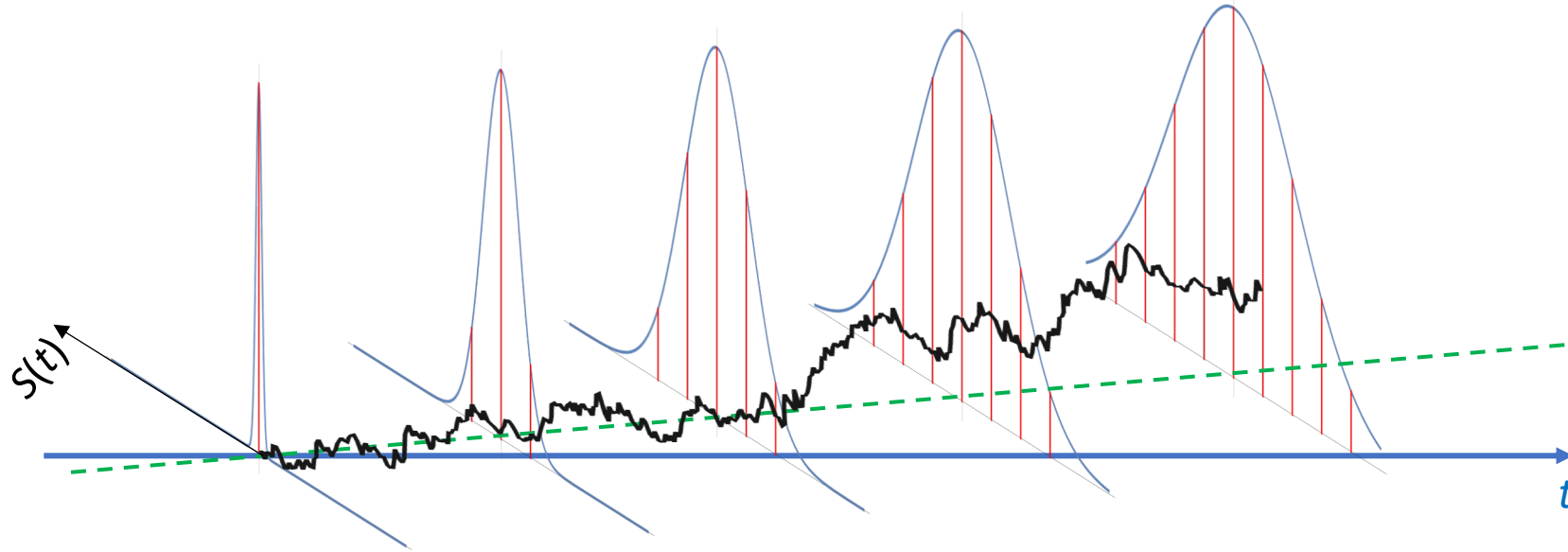
❖ The **price change** (e.g.  $S(t') - S(t)$  or  $S(t') / S(t)$ ) from time  $t$  to  $t'$  is also a **random variable** (usually with the same **PDF**)

- ❑ E.g. in the **Binomial** process in each time step the price  $S(t)_i$  can change only to  $S(t+1)_{i\pm 1}$
- ❑ Therefore: the random variables in times  $t$  and  $t' > t$  are usually (but not always) correlated, the observed random variable **REALIZATION**  $S(t')_i$  in time  $t'$  depends on the realization  $S(t)_i$  in time  $t$



# Stochastic processes (short introduction)

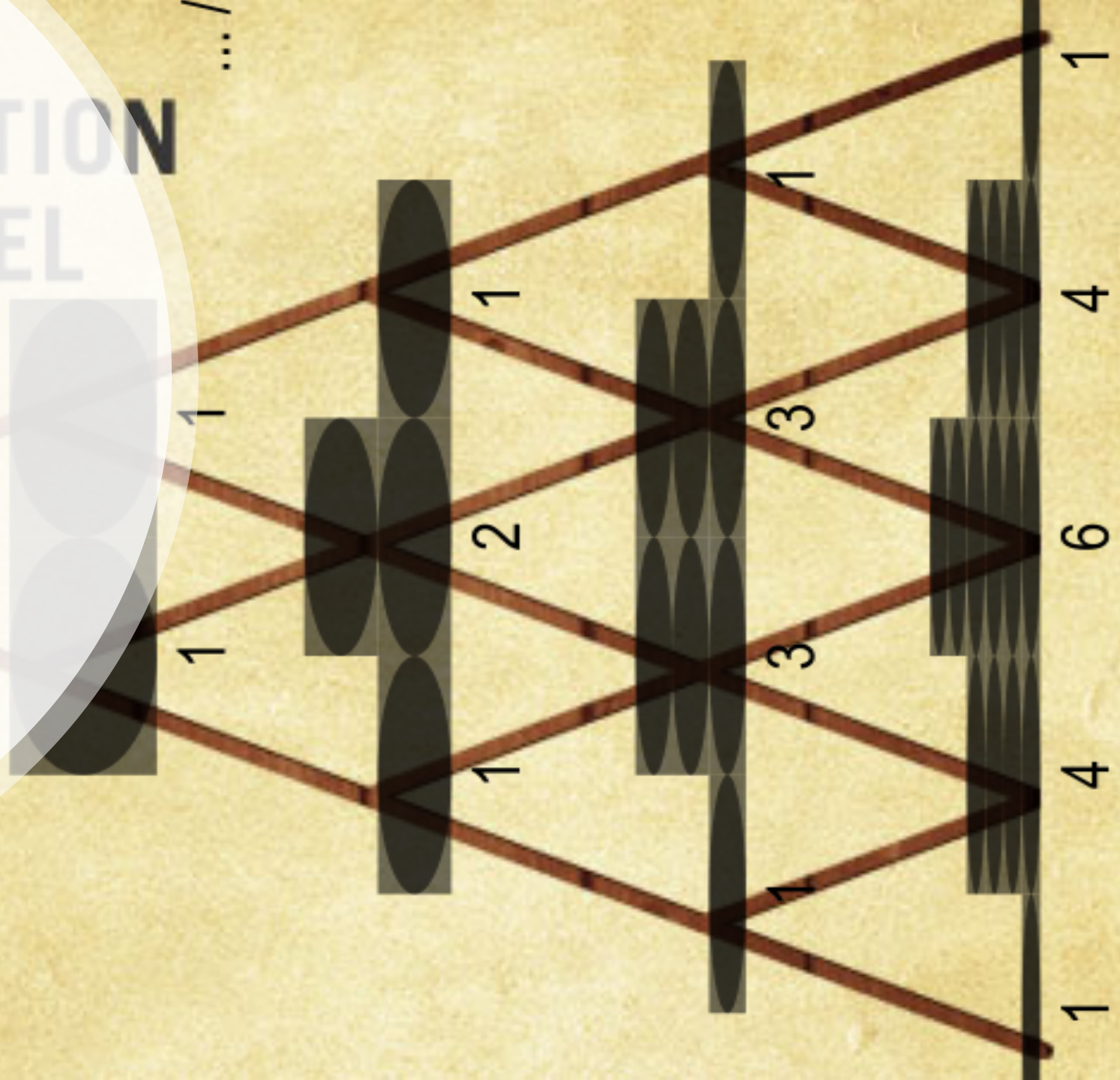
❖ A **stochastic process** is a set (family) of **random variables indexed by time**:  $\{S(t)\}_t$



- ❖ Sometimes some **additional regularity / continuity / smoothness** conditions apply
- E.g. in the **Wiener** process (Brownian motion)  $\{S(t)\}_t$  should have **continuous trajectories**, i.e. observed realizations  $S(t)$  should be continuous functions of time  $t$  (However, interestingly: they are nowhere differentiable !)

# Option pricing in Binomial Model

- ❖ Stochastic processes
- ❖ **One-step Binomial Model**
- ❖ Multi-step Binomial Model
- ❖ “Analytic” formula for European options
- ❖ Risk-neutral / martingale pricing





# Binomial model: options notation

❖ To remind the options **notation**:

❑  $S(t)$  – the „underlying asset” (e.g. **S**hare) price in time  $t$ ,  $S(0) = S$

❑  $X$  - the e**X**ercise price\*

❑  $T$  - expiration date

❑  $c(t)$  - **E**uropean “**c**all” option value/price at time  $t$ ,  $c(0) = c$  is the premium paid in  $t=0$

❑  $p(t)$  - **E**uropean “**p**ut” option value/price,  $p(0) = p$

❑  $C(t)$  - **A**merican “**C**all” option value/price,  $C(0) = C$

❑  $P(t)$  - **A**merican “**P**ut” option value/price  $P(0) = P$

Exercised only  
at EXPIRATION

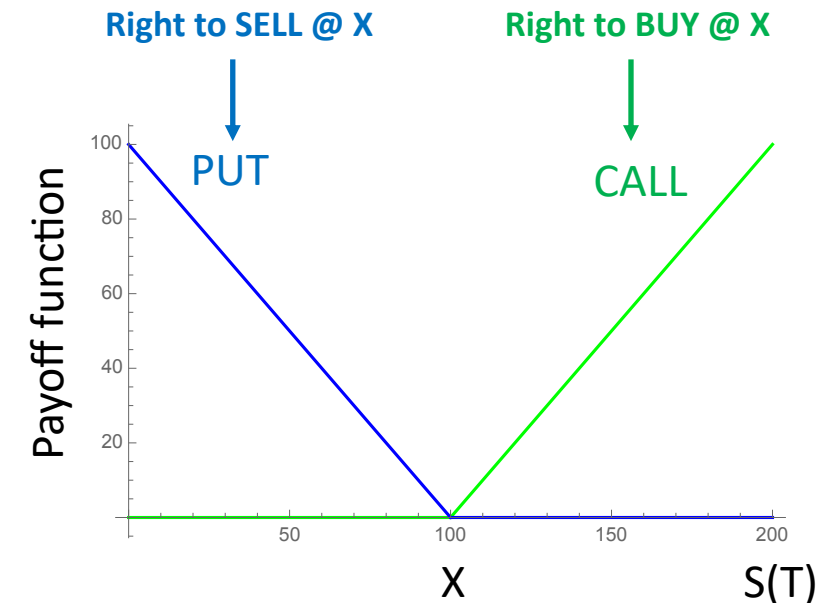
CAN be  
EXERCISED EARLY

❖ The **value of the option at expiration**  $T$  is:

❑  $c(T) = C(T) = \max(S(T) - X; 0)$

❑  $p(T) = P(T) = \max(X - S(T); 0)$

❖ These are also “**payoff functions**” (always paid by option SELLER to option BUYER\*\*) of options settled in cash (i.e. with no physical delivery)



\* In the literature the e**X**ercise price is sometimes denoted: **K** (stri**K**e price)

\*\***Assymmetric risk !** But in exchange of option price (premium) which is always paid by the option BUYER to the option SELLER !

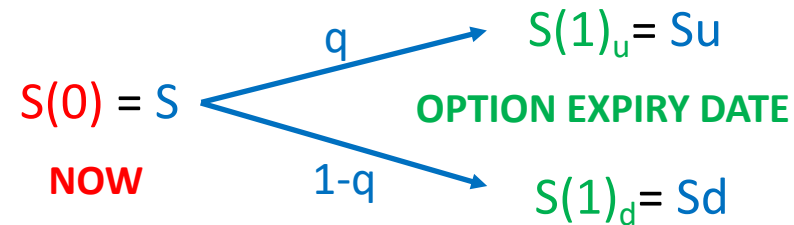
# Binomial model (one step)

- ❖ Let us start with a simple\* **Binomial model** introduced in late 70's by J. **Cox**, S. **Ross** and M. **Rubinstein** (sometimes called the **CRR** model)
- ❖ Let us assume that the **European call option** expires in **N=1 time steps** (time is counted in discrete steps)
- ❖ The **current share price** is  $S(0) = S$  with probability one (this is the observed current market price in  $t=0$ )
- ❖ In next step, i.e. **on the option expiry**, the share price can either change to  $S(1)_u = S \times u$  **with probability  $q$**  or to  $S(1)_d = S \times d$  **with probability  $(1-q)$**  (one assumes:  $u > d$ )
- ❖ By **adjusting parameters  $u$ ,  $d$  and  $q$**  one can manipulate the future (@ option's expiry):
  - **Expected** share price:  $E(S(1)) \equiv \langle S(1) \rangle = q S(1)_u + (1-q) S(1)_d = S(qu + (1-q)d)$
  - **and Standard Deviation**:  $\sigma(S(1)) \equiv (\langle S(1)^2 \rangle - \langle S(1) \rangle^2)^{1/2} = S q^{1/2} (1-q)^{1/2} (u-d)$

\* Although the **Binomial** model is very simple it has all properties of other (more complicated) pricing models, and, as will be shown in **Lecture 7**, it tends to the, so called, **Black-Scholes** model if the number of discrete time steps:  $N \rightarrow \infty$ .

# Binomial model (one step)

❖ Binomial model **parameters**:  $S$ ,  $u$ ,  $d$  (and  $q$ ) are assumed to be known



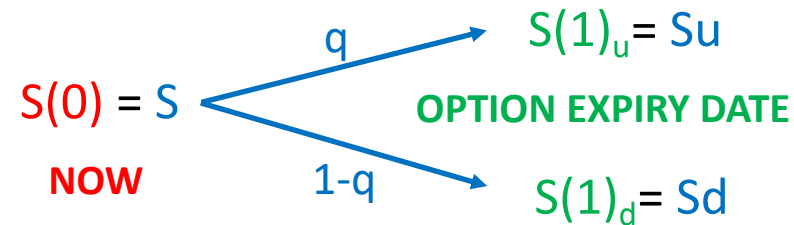
❖ One **additionally assumes** the following:

- ❑ One can **buy / sell any non-integer fraction** of shares (not a problem for “big” portfolios)
- ❑ **Short selling** is possible (one can have negative number of shares)
- ❑ One can **borrow/deposit** any cash amount with (the same) effective “**RISK-FREE**”\* **interest rate  $R\%$**   
**NOTE:** here interest rate  $R\%$  is the **yield per discrete time period**, one does not have to adjust it by any DayCountFactor, i.e. if one borrows/deposits  $B$ , the paid/received interest is simply:  $INT(1)=B \times R$
- ❑ No transaction costs, margins, taxes, ...

\*Here „**RISK-FREE**” refers to market (price) risk. Shares are „risky” as their prices are random variables with a priori unknown (realized) returns. Deposits / loans are „risk-free” as they have a priori fixed/known returns (yields)  $R\%$

# Binomial model (one step)

❖ Binomial model **parameters**:  $S$ ,  $u$ ,  $d$  (and  $q$ ) are assumed to be known

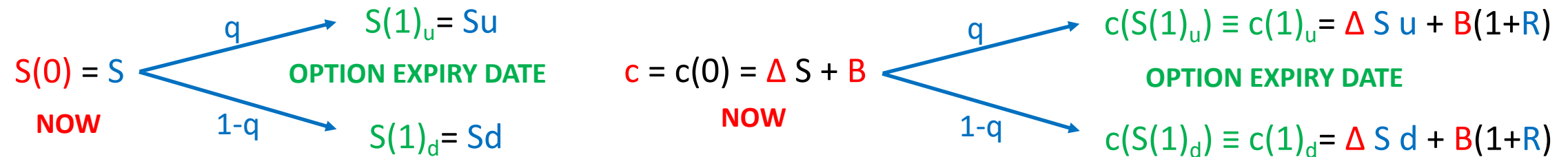


❖ Assume that the **European CALL option can be REPLICATED by**:

- **BUYING** (in  $t=0$ )  $\Delta$  **shares** ( $\Delta$  is a real number from  $[-1 ; 1]$ , when  $\Delta < 0$  one sells shares short)
- **And making a deposit of  $B$**  (when  $B < 0$  one takes a loan)



# Binomial model (one step)



❖ Assume that the **European CALL option can be REPLICATED** by:

- **BUYING** (in  $t=0$ )  $\Delta$  **shares** ( $\Delta$  is a real number from  $[-1; 1]$ , when  $\Delta < 0$  one sells shares (short))
- **And making a deposit of  $B$**  (when  $B < 0$  one takes a loan)

□ Therefore **NOW** (in  $t=0$ ) one has:

$$c \equiv c(0) = \Delta \times S + B$$

□ And **on the option EXPIRY** (in  $t=1$  time steps):

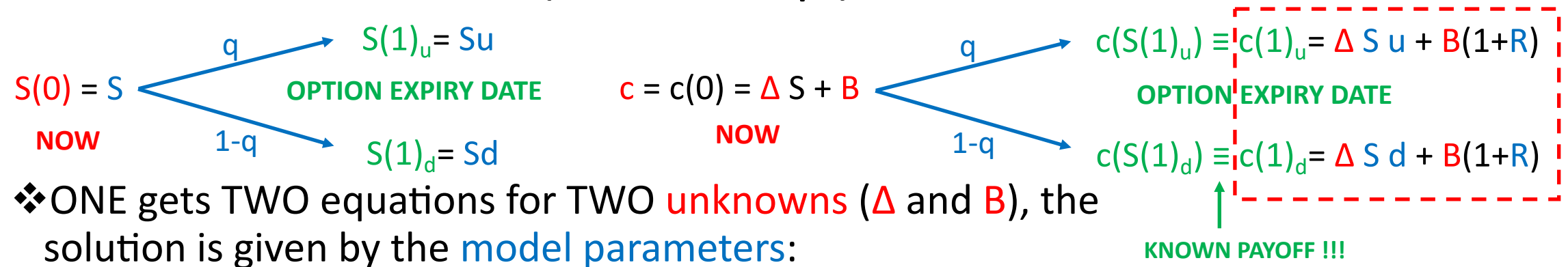
$$c(1)_u = \Delta S u + B(1+R) \text{ with probability } q \quad \text{or} \quad c(1)_d = \Delta S d + B(1+R) \text{ with probability } (1-q)$$

□ But **on the EXPIRY** date, the (random) option price is simply given by its **PAYOFF function**:

$$c(1) \equiv c(S(1)) = \max(S(1) - X; 0)$$

□ Thus both:  $c(1)_u = c(S(1)_u) = \max(S u - X; 0)$  and  $c(1)_d = c(S(1)_d) = \max(S d - X; 0)$  are known (can be computed from the model parameters:  $S$ ,  $u$  &  $d$ ) !!!

# Binomial model (one step)



❖ ONE gets TWO equations for TWO **unknowns** ( $\Delta$  and  $B$ ), the solution is given by the **model parameters**:

$$\begin{cases} c(1)_u = \Delta S u + B(1+R) \\ c(1)_d = \Delta S d + B(1+R) \end{cases} \quad \rightarrow \quad \begin{cases} \Delta = \frac{c(1)_u - c(1)_d}{S(u - d)} \\ B = -\frac{d c(1)_u - u c(1)_d}{(u - d)(1 + R)} \end{cases}$$

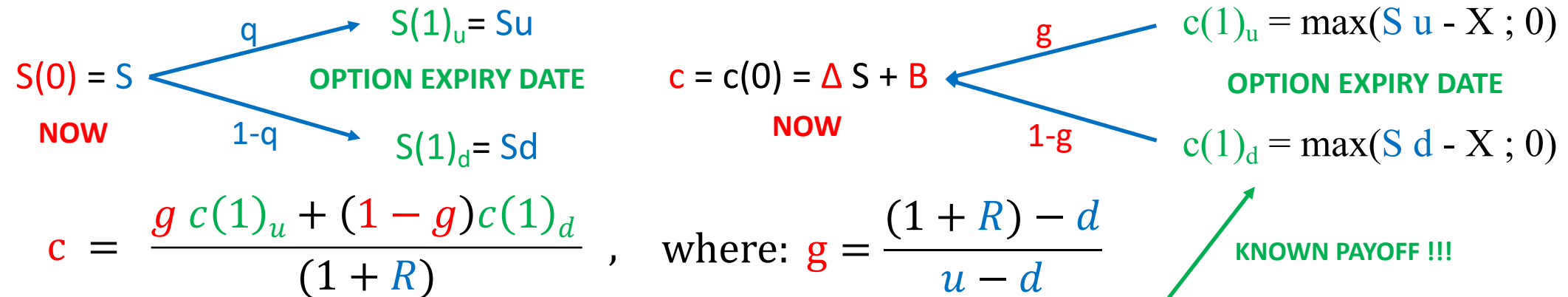
$\begin{cases} c(1)_u = \max(S u - X; 0) \\ c(1)_d = \max(S d - X; 0) \end{cases}$

❖ Knowing  $\Delta$  and  $B$  one computes (current) option “**fair**” **price** (based on arbitrage-free arguments: the option can be replicated so its price must equal the portfolio price):

$$c \equiv c(0) = \Delta \times S + B$$

$$c = \frac{g c(1)_u + (1 - g) c(1)_d}{(1 + R)}, \quad \text{where: } g = \frac{(1 + R) - d}{u - d}$$

# Binomial model (one step)

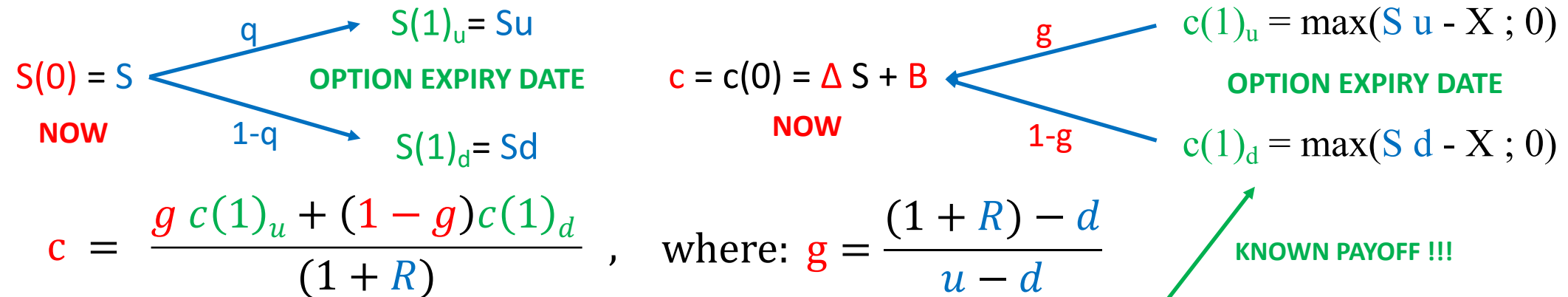


❖ The **option price** depends on:

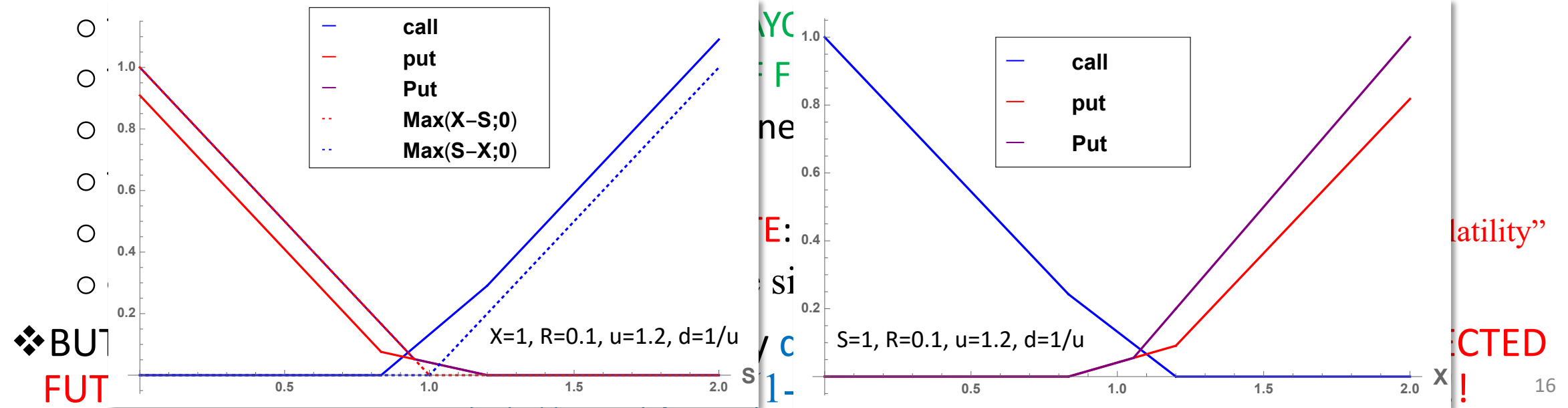
- the current share price:  $S$  (through the **PAYOFF FUNCTION**)
- the eXercise price:  $X$  (through the **PAYOFF FUNCTION**)
- NOTE: using other **PAYOFF FUNCTIONS** one can also price PUTs and other options
- the "RISK FREE" interest rate  $R\%$
- Binominal model parameters:  $u$  &  $d$  (NOTE:  $(u-d) \propto \sigma(S(1)) = S q^{1/2} (1-q)^{1/2} (u-d) \leftarrow$  "Volatility")
- option type, i.e. for **American options** one simply sets:  $C = \max(S - X; c)$

❖ BUT it **DOES NOT DEPEND ON** probability  $q$ , i.e. DOES NOT DEPEND ON **THE EXPECTED FUTURE SHARE PRICE:  $E(S(1)) = S(q u + (1-q) d)$**  (the **expected return** on shares) !!!

# Binomial model (one step)



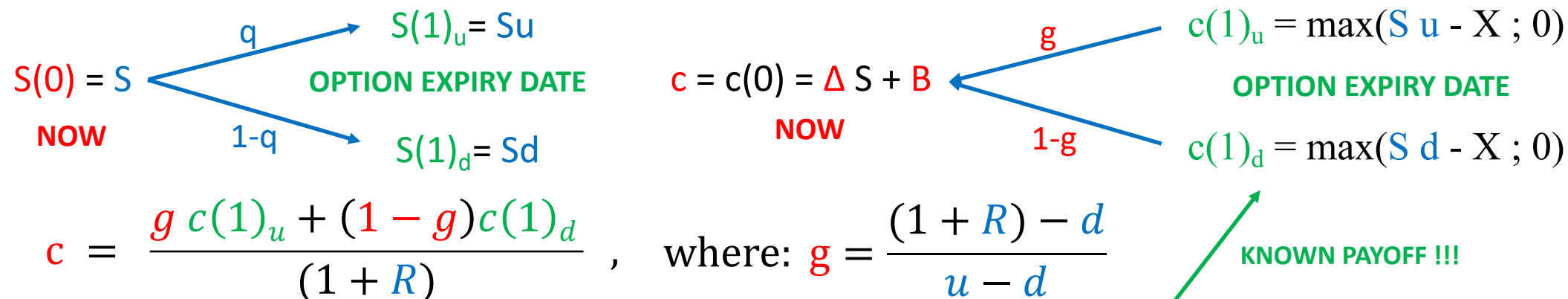
❖ The **option price** depends on:



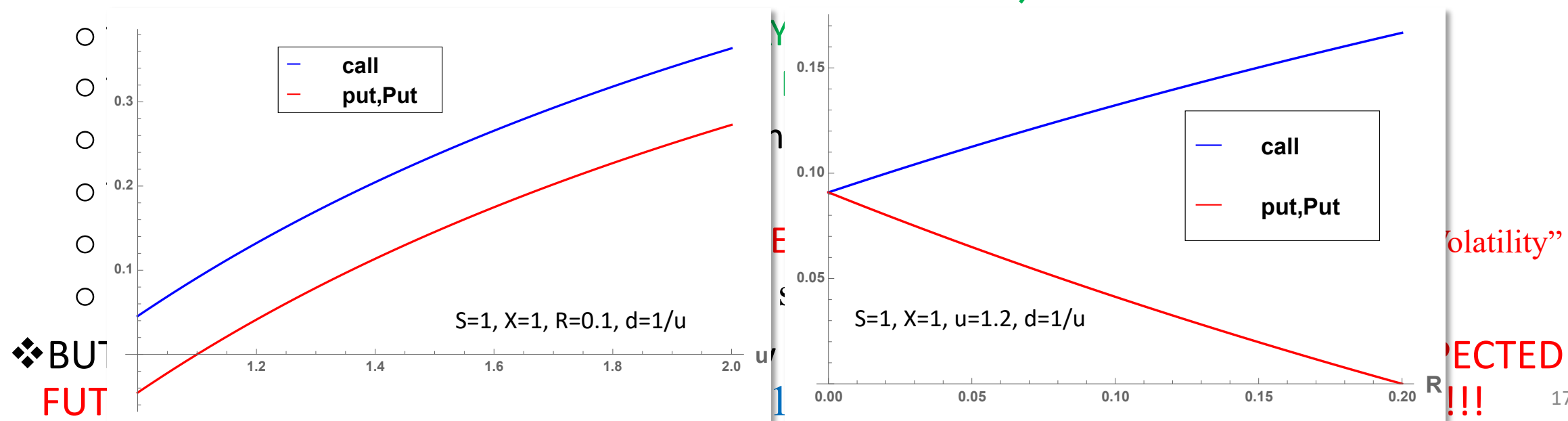
❖ BUT  
FUT



# Binomial model (one step)

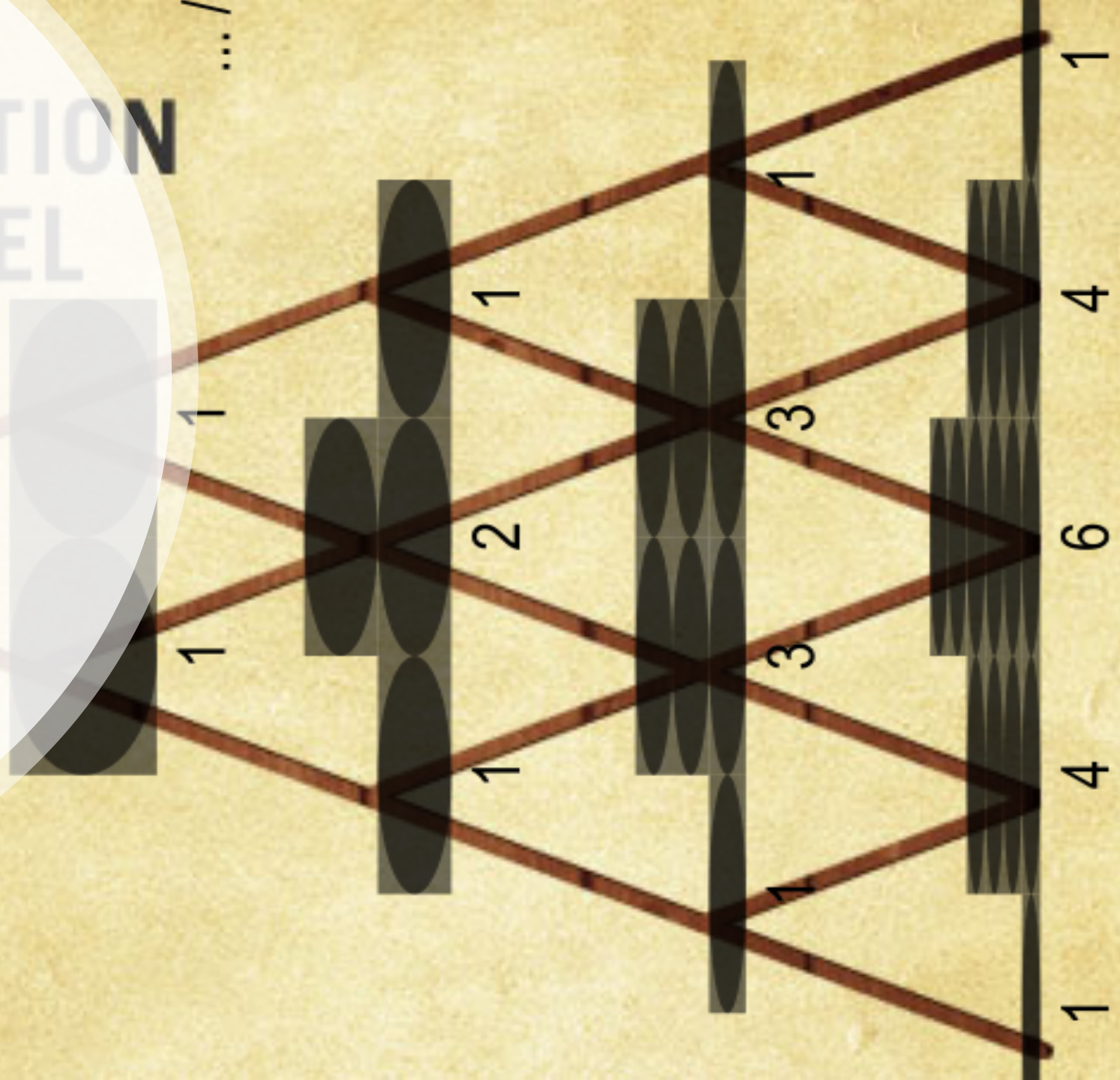


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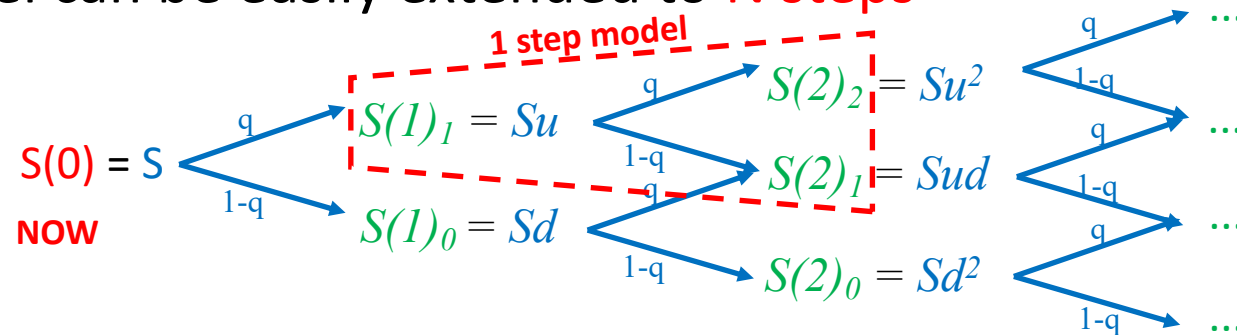
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# Binomial model (N steps)

❖ The model can be easily extended to **N steps**



OPTION EXPIRY DATE ( $n=N$ )

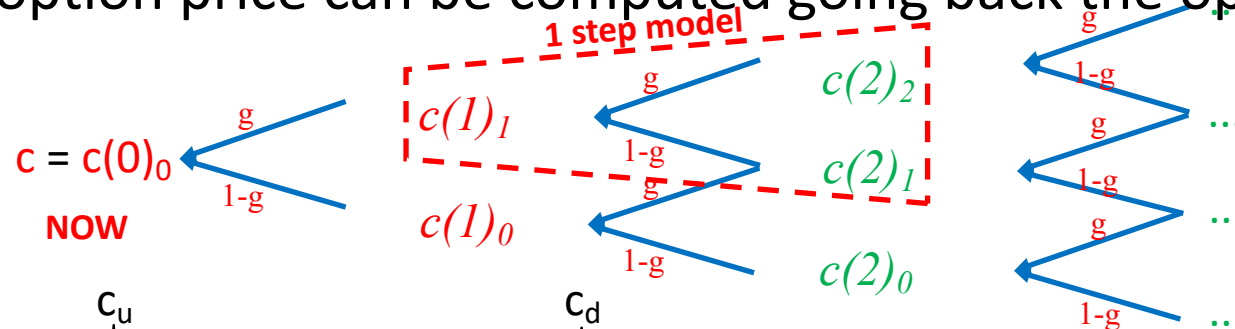
$$S(N)_j = S u^j d^{N-j}$$

❖ After  $n$  steps the **share price**:  $S(n)_j = S u^j d^{n-j}$  has **Binomial** probability distribution

How many paths  
lead to  $j$ -th price

$$\mathcal{P}_q(n)_j = \binom{n}{j} q^j (1-q)^{n-j}, \quad j = 0, 1, \dots, n$$

❖ And the option price can be computed going back the option Binomial tree



KNOWN PAYOFF !!! on the  
OPTION EXPIRY DATE ( $n=N$ )

$$c(N)_j = \max(S u^j d^{N-j} - X; 0)$$

$$c(n)_j = \frac{\overbrace{g c(n+1)_{j+1}}^{C_u} + \underbrace{(1-g) c(n+1)_j}_{C_d}}{(1+R)}$$

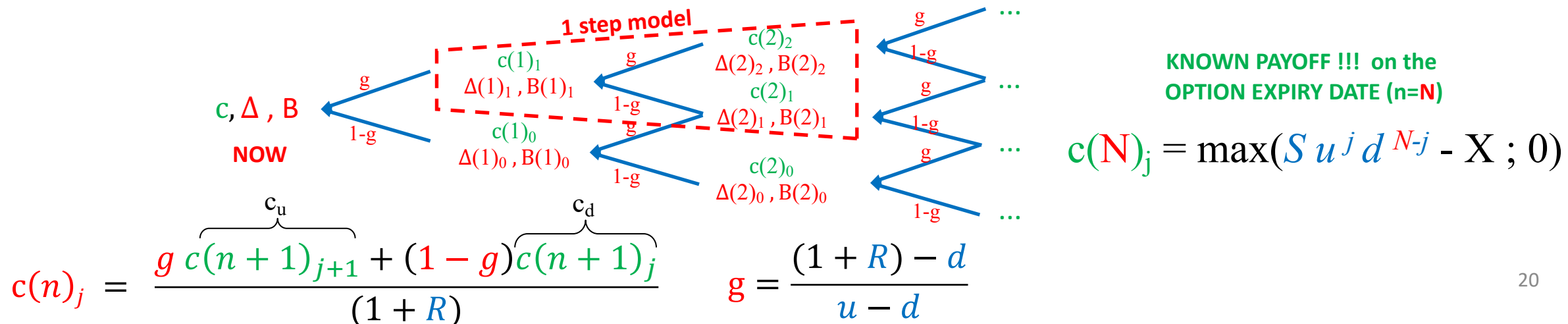
$$g = \frac{(1+R) - d}{u - d}$$

# Binomial model (N steps)

- ❖ After  $n$  steps the **share price**:  $S(n)_j = S u^j d^{n-j}$  has **Binomial** probability distribution
- ❖ For each step  $n$  one can easily compute the number of shares  $\Delta(n)_j$  and the value of deposit/loan  $B(n)_j$  used to create the replicating portfolio:  $c(n)_j = \Delta(n)_j \times S(n)_j + B(n)_j$

$$\Delta(n)_j = \frac{\overbrace{c(n+1)_{j+1}}^{c_u} - \overbrace{c(n+1)_j}^{c_d}}{S(u-d)} \quad B(n)_j = -\frac{d \overbrace{c(n+1)_{j+1}}^{c_u} - u \overbrace{c(n+1)_j}^{c_d}}{(u-d)(1+R)}$$

- ❖ Note that now these depend on the evolution of the share price  $S(n)_j$  and change in each step, thus the **replicating portfolio has to be constantly adjusted**





# Binomial model (N steps): Example

❖ Assume:  $u=1.2$ ,  $d=1/u$ ,  $R=0.1$ ,  $S=1$  and consider **European call** with  $X = 1$ , expiring in  $N = 3$  steps

$$S(0) = 1$$

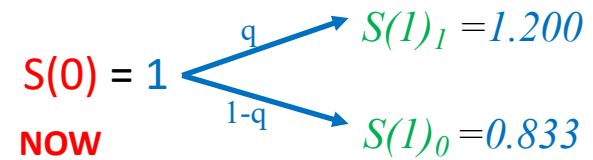
**NOW**

$$c = ???$$

**NOW**

# Binomial model (N steps): Example

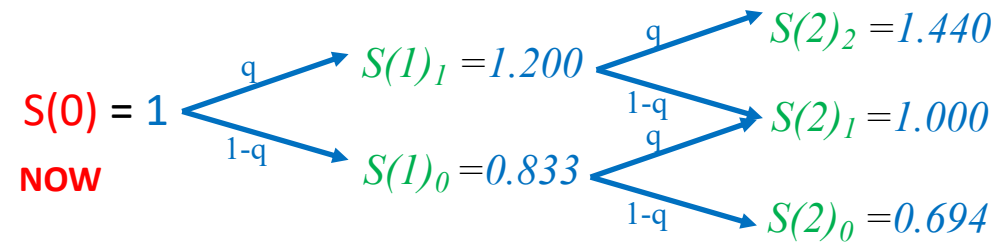
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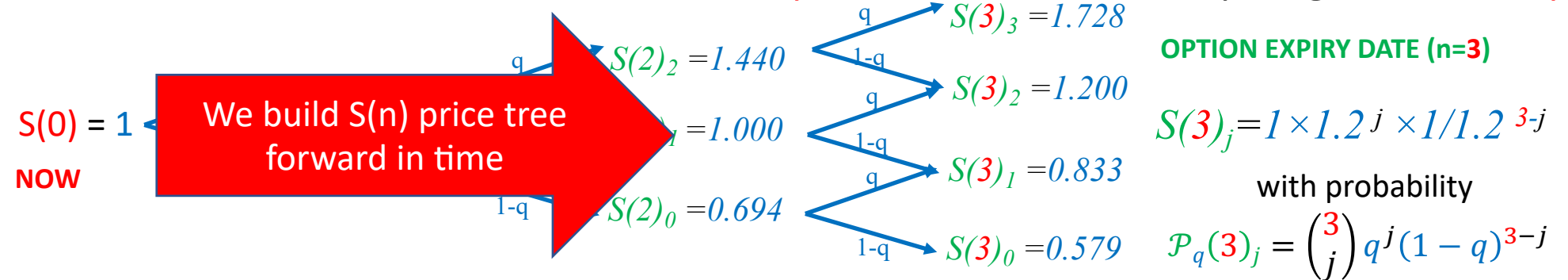


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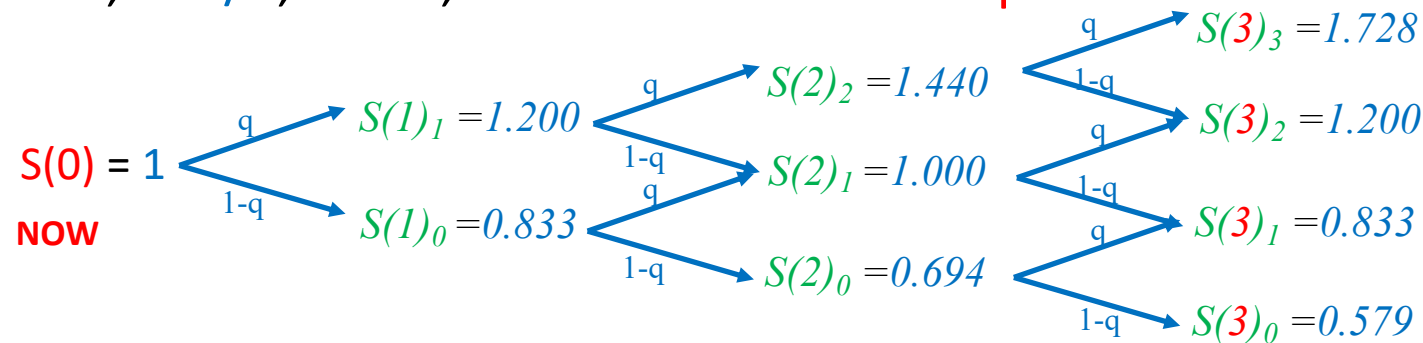
$c = ???$

NOW



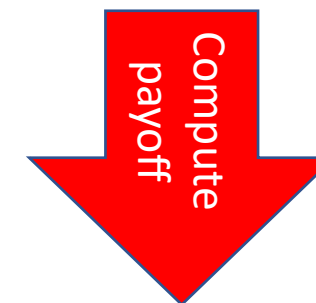
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**OPTION EXPIRY DATE (n=3)**

$$S(3)_j = 1 \times 1.2^j \times 1/1.2^{3-j}$$



$c = ???$   
**NOW**

$c(3)_3 = 0.728$  **KNOWN PAYOFF !!! on the**  
**OPTION EXPIRY DATE (n=3)**

$$c(3)_2 = 0.200$$

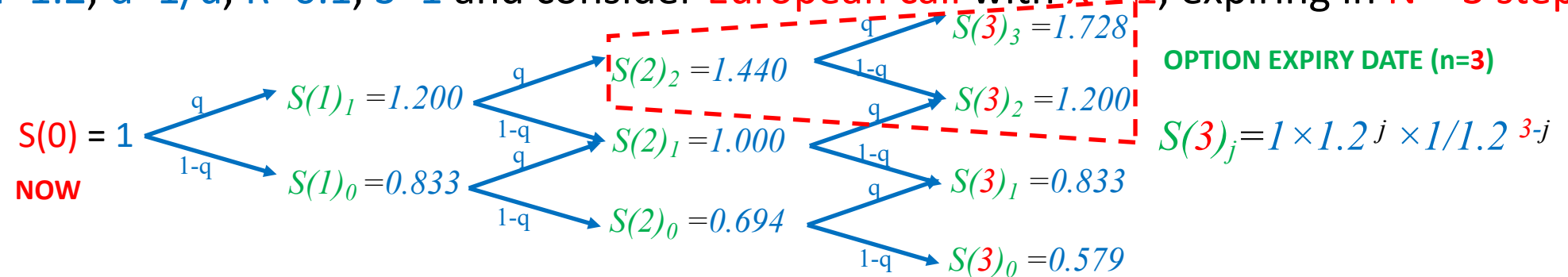
$$c(3)_j = \max(S(3)_j - X; 0) = \max(1 \times 1.2^j \times 1/1.2^{3-j} - 1; 0)$$

$$c(3)_1 = 0$$

$$c(3)_0 = 0$$

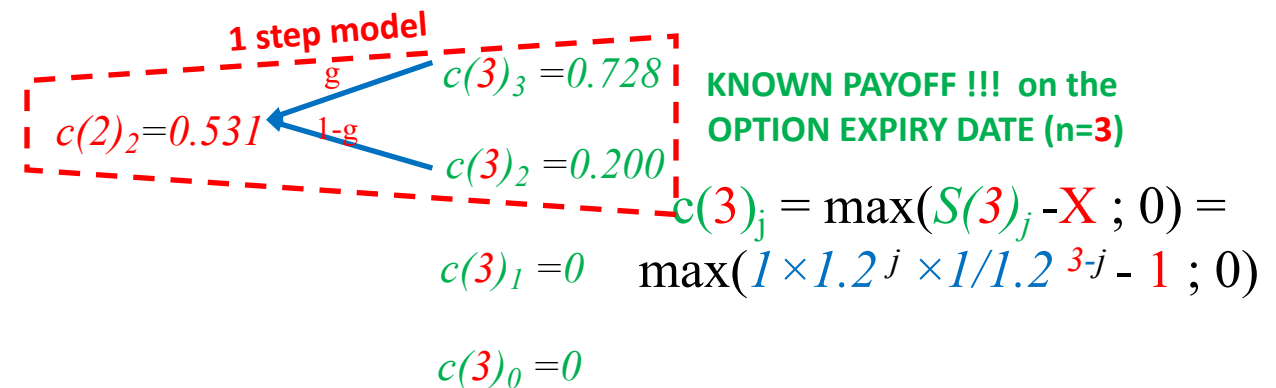
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$c = ???$   
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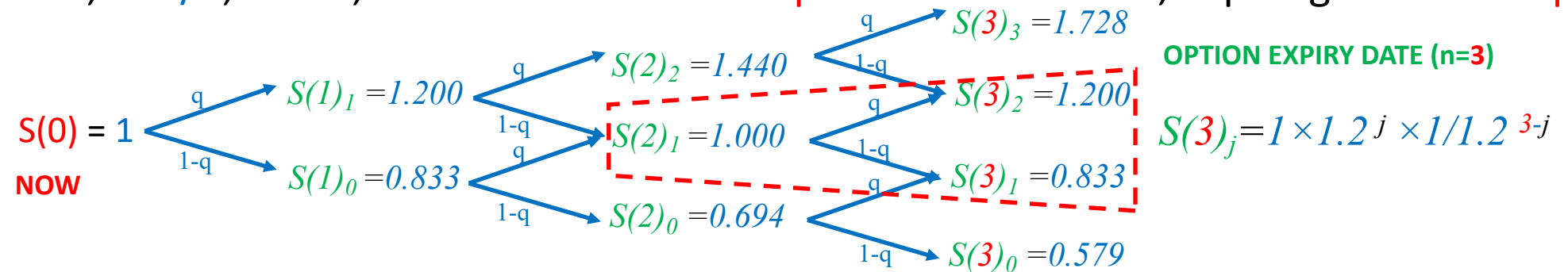
$$c(n)_j = \frac{\overbrace{g c(n+1)_{j+1}}^{c_u} + (1-g) \overbrace{c(n+1)_j}^{c_d}}{(1+R)}$$



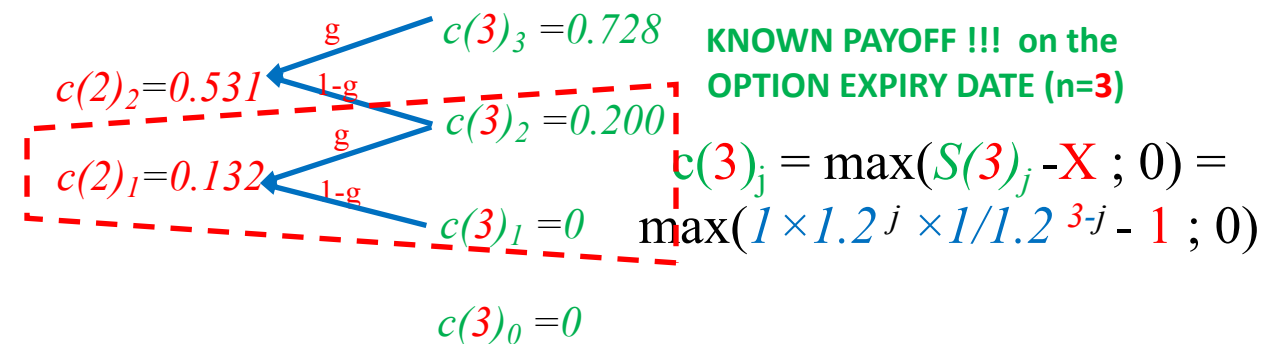
$$g = \frac{(1+R) - d}{u - d} = 0.7273$$

# Binomial model (N steps): Example

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**NOW**

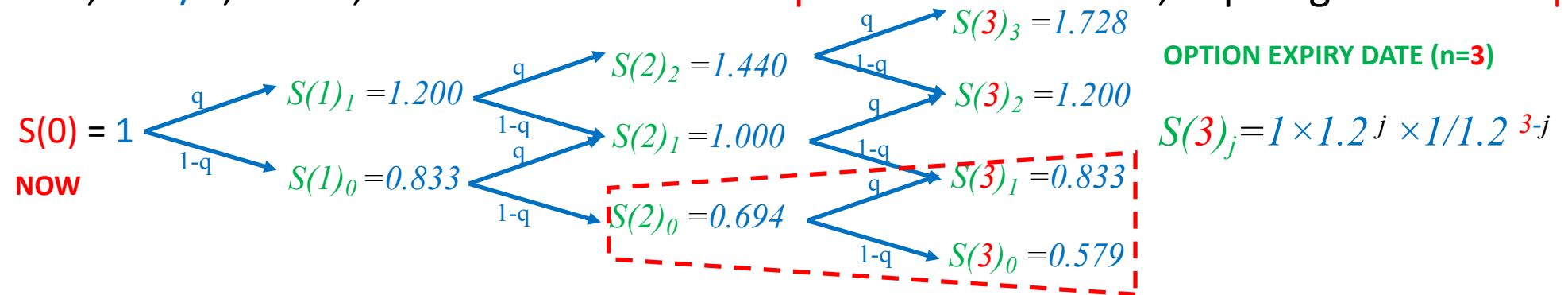


$$c(n)_j = \frac{\overbrace{g c(n+1)_{j+1}}^{c_u} + (1-g) \overbrace{c(n+1)_j}^{c_d}}{(1+R)}$$

$$g = \frac{(1+R) - d}{u - d} = 0.7273$$

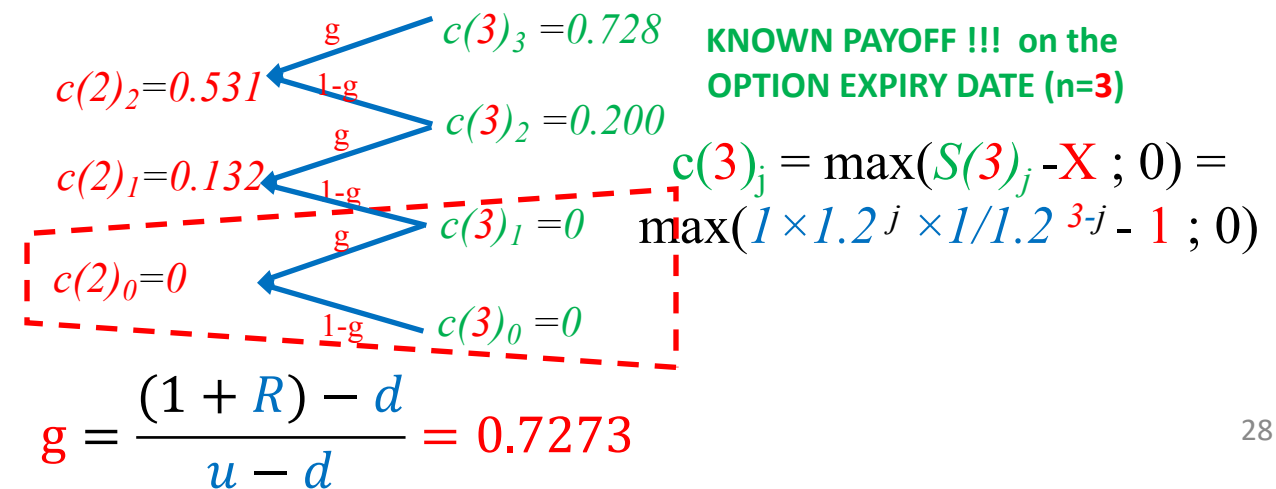
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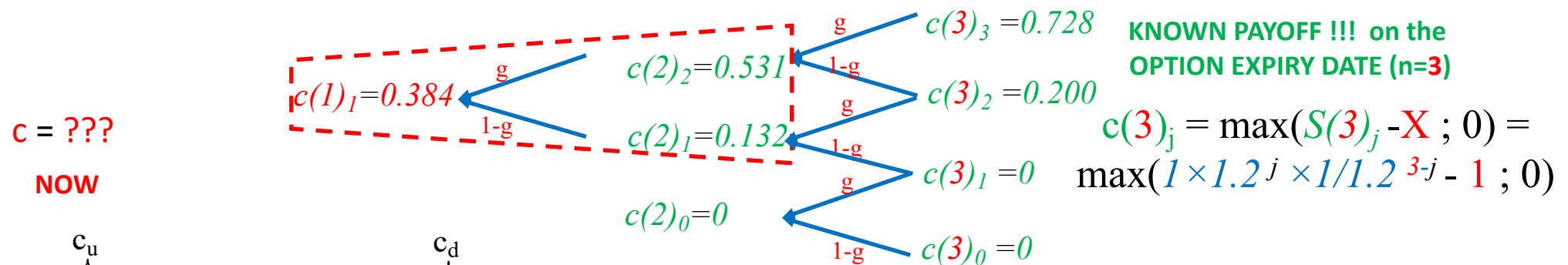
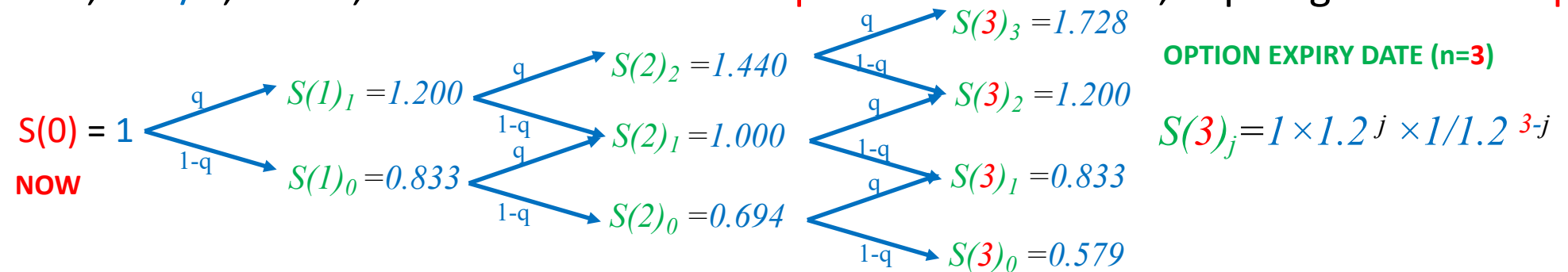
$$c(n)_j = \frac{\overbrace{g c(n+1)_{j+1}}^{c_u} + (1-g) \overbrace{c(n+1)_j}^{c_d}}{(1+R)}$$



$$g = \frac{(1+R) - d}{u - d} = 0.7273$$

# Binomial model (N steps): Example

❖ Assume:  $u=1.2$ ,  $d=1/u$ ,  $R=0.1$ ,  $S=1$  and consider **European call** with  $X = 1$ , expiring in  $N = 3$  steps

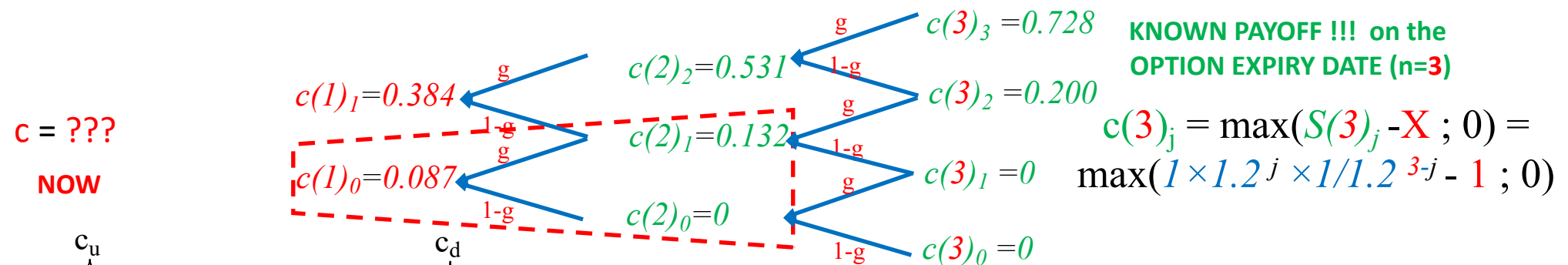
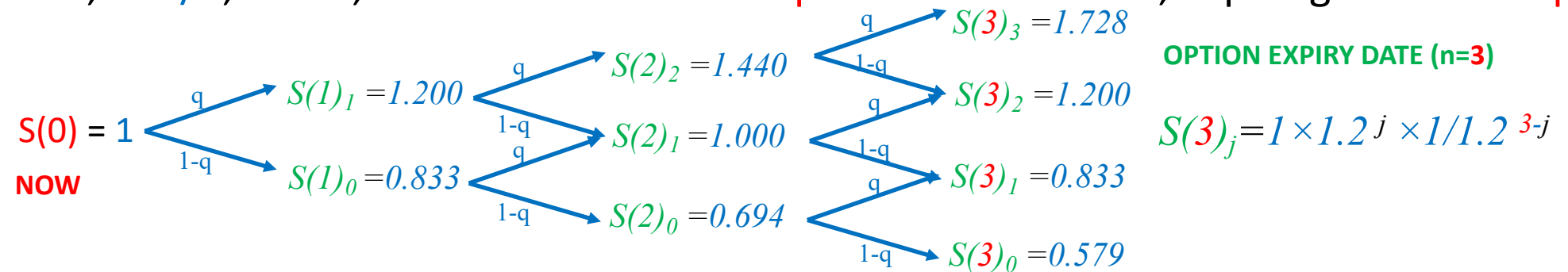


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$$g = \frac{(1+R) - d}{u - d} = 0.7273$$

# Binomial model (N steps): Example

❖ Assume:  $u=1.2$ ,  $d=1/u$ ,  $R=0.1$ ,  $S=1$  and consider **European call** with  $X = 1$ , expiring in  $N = 3$  steps



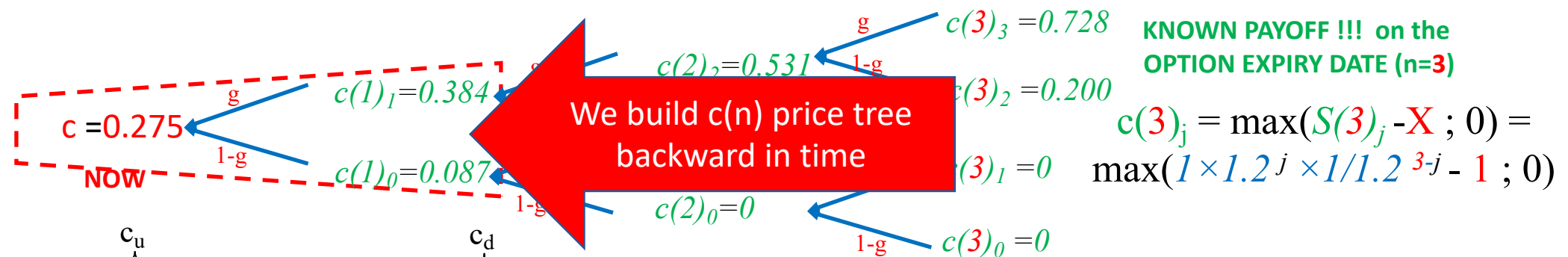
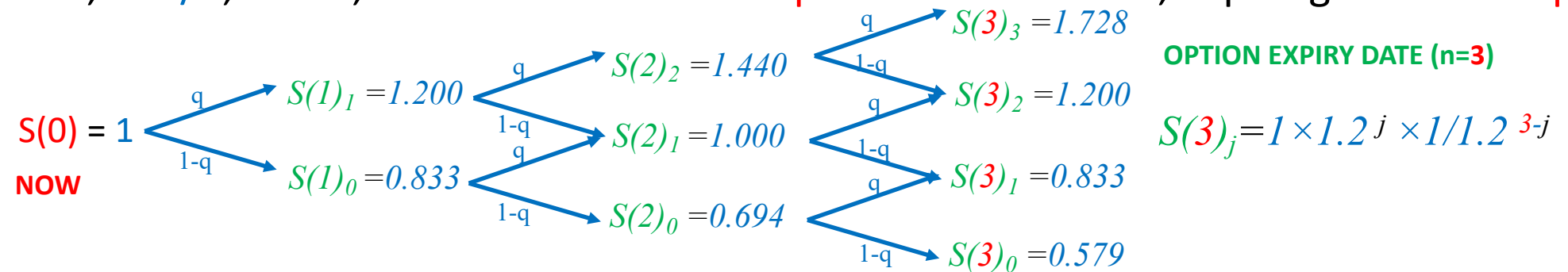
$$c(n)_j = \frac{\overbrace{g c(n+1)_{j+1}}^{c_u} + \overbrace{(1-g)c(n+1)_j}^{c_d}}{(1+R)}$$

$$g = \frac{(1+R) - d}{u - d} = 0.7273$$



# Binomial model (N steps): Example

❖ Assume:  $u=1.2$ ,  $d=1/u$ ,  $R=0.1$ ,  $S=1$  and consider **European call** with  $X = 1$ , expiring in  $N = 3$  steps

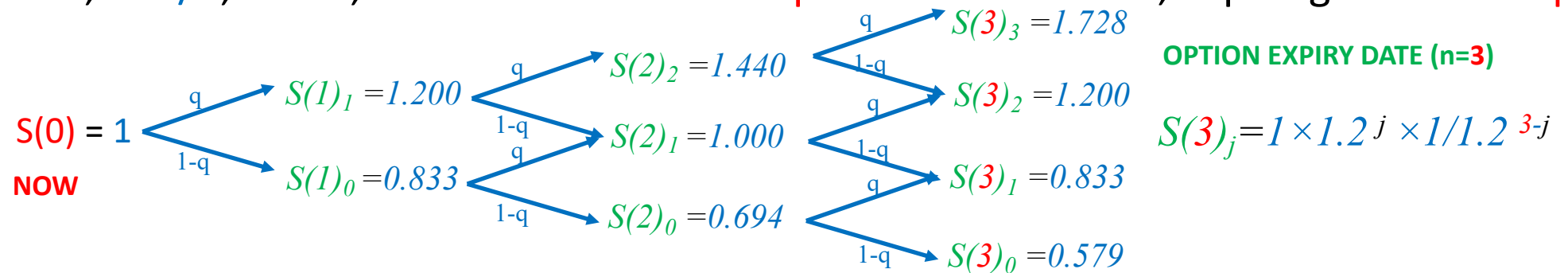


$$c(n)_j = \frac{\overbrace{g c(n+1)_{j+1}}^{c_u} + \overbrace{(1-g)c(n+1)_j}^{c_d}}{(1+R)}$$

$$g = \frac{(1+R) - d}{u - d} = 0.7273$$

# Binomial model (N steps): Example

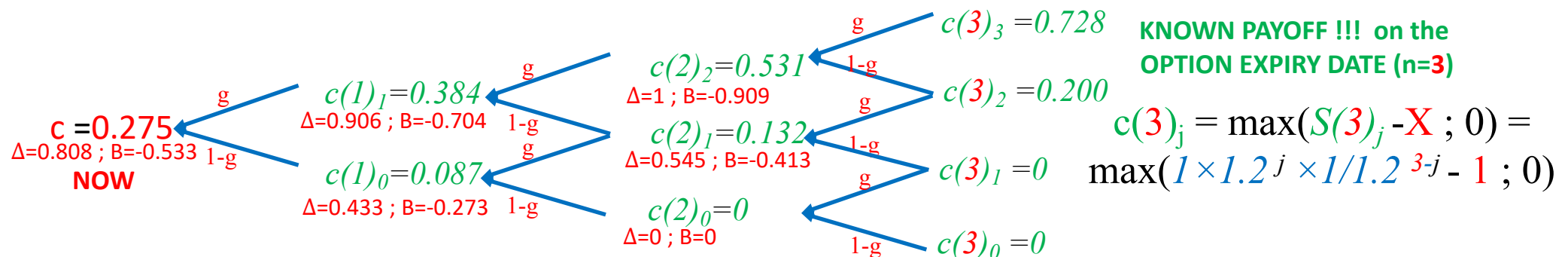
❖ Assume:  $u=1.2$ ,  $d=1/u$ ,  $R=0.1$ ,  $S=1$  and consider **European call** with  $X = 1$ , expiring in  $N = 3$  steps



❖ Performing the same way one can also compute  $\Delta(n)_j$  and  $B(n)_j$

❑ NOTE that for the **call** option  $\Delta(n)_j \geq 0$  and  $B(n)_j \leq 0$ , so to replicate the option we BUY shares and take a LOAN

❑ AND they vary in each time step so the **REPLICATING PORTFOLIO** must be **DYNAMICALLY ADJUSTED !!!**



$$\Delta(n)_j = \frac{c(n+1)_{j+1} - c(n+1)_j}{S(n)_j(u - d)}$$

$$B(n)_j = -\frac{d c(n+1)_{j+1} - u c(n+1)_j}{(u - d)(1 + R)}$$

# Binomial model: Dynamical hedging/arbitrage

❖ We (NOW:  $n=0$ ) **SELL the call option** from the previous Example

$$S(0) = 1$$

**NOW**

❖ In order to **HEDGE the SHORT OPTION** position we (NOW) **BUY THE REPLICATING PORTFOLIO**:  $\Delta \times S + B$ , i.e. we BUY  $\Delta = 0.808$  SHARES @  $S=1$  and take a LOAN of  $B = -0.533$  @  $R=10\%$  (per step)

$$c = 0.275$$

$$\Delta = 0.808; B = -0.533$$

**NOW**

**CURRENT POSITIONS ( $n=0$ ):**

**SHORT call:  $-c$**

**LONG shares:  $+0.808 S$**

**Loan:  $B = -0.533$**

**CURRENT NET WORTH\* = 0 !**

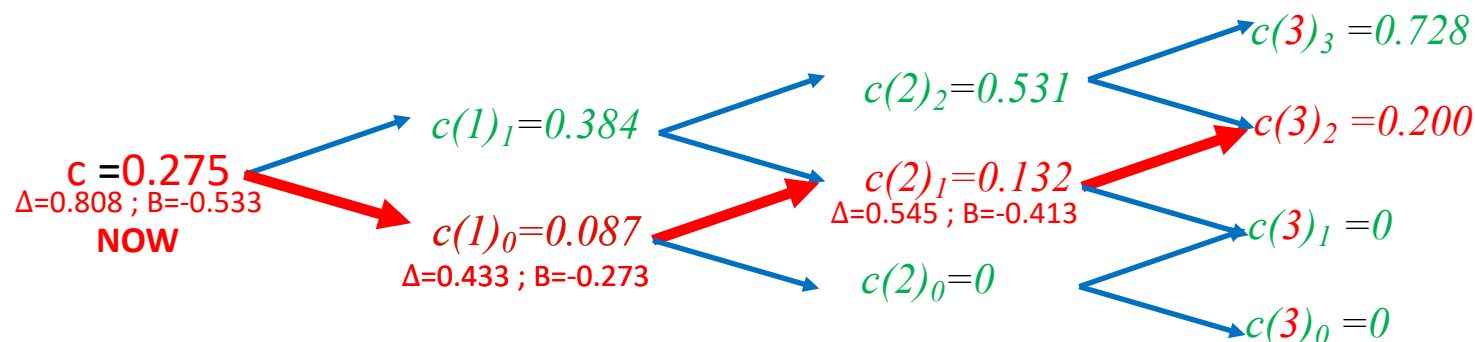
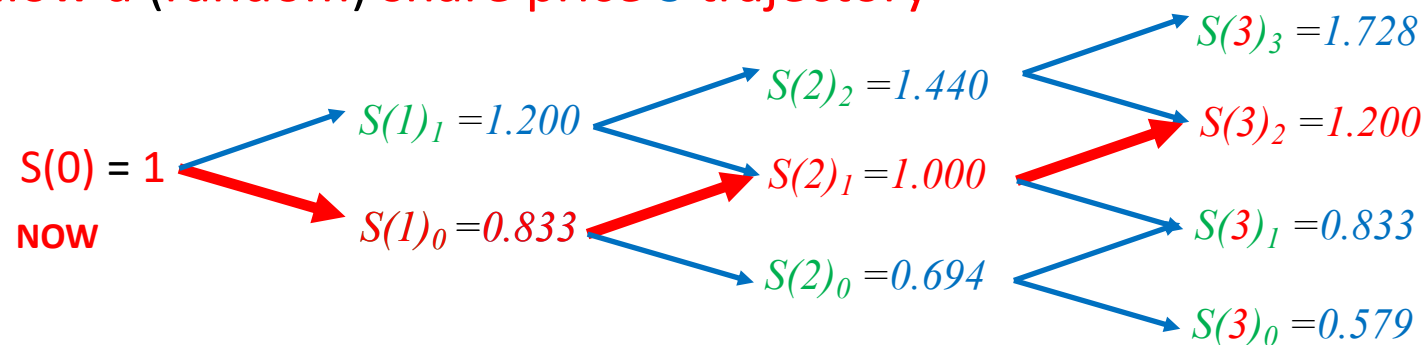
\*IF OPTION PRICES ARE FAIR:  $c = \Delta \times S + B$

\*\*If one closes all positions one gets  $CF=0$

❖ NOTE that the current  $CF(0) = 0$ , as  $c = \Delta \times S + B$ , so we BUY the shares ( $CF = -0.808 \times 1$ ) from the INFLOWS OF SELLING OPTION ( $CF = +0.275$ ) and from the LOAN ( $CF = +0.533$ ) 33

# Binomial model: Dynamical hedging/arbitrage

❖ We will follow a (random) share price  $S$  trajectory



**CURRENT POSITIONS ( $n=0$ ):**

**SHORT** call:  $-c$

**LONG** shares:  $+0.808 S$

**Loan:**  $B = -0.533$

**CURRENT NET WORTH\* = 0 !**

\*IF OPTION PRICES ARE FAIR:  $c = \Delta \times S + B$

\*\*If one closes all positions one gets  $CF=0$

# Binomial model: Dynamical hedging/arbitrage

❖ We follow a (random) share price  $S$  trajectory: in NEXT STEP ( $n=1$ ) the share price ↓

$$\begin{array}{l} S(0) = 1 \\ \text{NOW} \end{array} \rightarrow S(1)_0 = 0.833$$

❖ We adjust THE REPLICATING PORTFOLIO, as  $S$  ↓ and thus  $\Delta$  ↓, we SELL:  $0.433 - 0.808 = -0.376$  SHARES @  $S = 0.833$  and, as  $B$  ↑, we REPAY OLD LOAN:  $-0.533 \times 1.1 = -0.586$  and we take a NEW (smaller) LOAN:  $-0.273$  @  $R = 10\%$

$$\begin{array}{l} c = 0.275 \\ \Delta = 0.808 ; B = -0.533 \\ \text{NOW} \end{array} \rightarrow \begin{array}{l} c(1)_0 = 0.087 \\ \Delta = 0.433 ; B = -0.273 \end{array}$$

CURRENT POSITIONS ( $n=1$ ):

SHORT call:  $-c$

LONG shares:  $+0.433 S$

Loan:  $B = -0.273$

CURRENT NET WORTH\* = 0 !

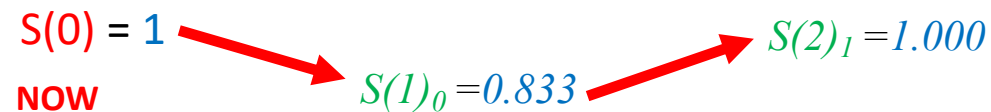
\*IF OPTION PRICES ARE FAIR:  $c = \Delta \times S + B$

\*\*If one closes all positions one gets  $CF = 0$

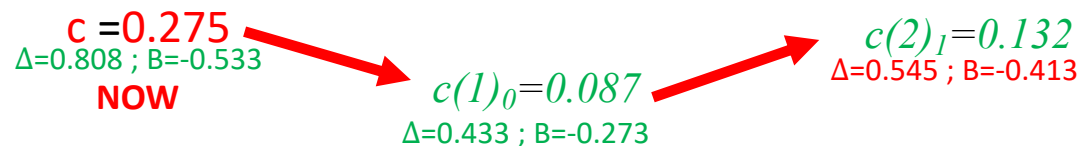
❖ NOTE that the current  $CF(1) = 0$ , as the INFLOWS from SELLING THE SHARES ( $CF = +0.376 \times 0.833 = +0.313$ ) and NEW LOAN ( $CF = +0.273$ ) exactly EQUAL OUTFLOWS FOR REPAYING THE OLD LOAN ( $CF = -0.586$ )

# Binomial model: Dynamical hedging/arbitrage

❖ We follow a (random) share price  $S$  trajectory: in NEXT STEP ( $n=2$ ) the share price  $\uparrow$



❖ We adjust **THE REPLICATING PORTFOLIO**, as  $S \uparrow$  and thus  $\Delta \uparrow$ , we BUY:  $0.545 - 0.433 = 0.113$  SHARES@ $S=1.000$  and, as  $B \downarrow$ , we REPAY OLD LOAN:  $-0.273 \times 1.1 = -0.300$  and we take a NEW (bigger) LOAN:  $-0.413$  @  $R=10\%$



**CURRENT POSITIONS ( $n=2$ ):**

**SHORT** call:  $-c$

**LONG** shares:  $+0.545 S$

**Loan:**  $B = -0.413$

**CURRENT NET WORTH\* = 0 !**

\*IF OPTION PRICES ARE FAIR:  $c = \Delta \times S + B$

\*\*If one closes all positions one gets  $CF=0$

❖ NOTE that the current  $CF(2) = 0$ , as the INFLOWS from NEW LOAN ( $CF = +0.413$ ) exactly EQUAL OUTFLOWS FOR BUYING THE SHARES ( $CF = -0.113 \times 1.000 = -0.113$ ) and FOR REPAYING THE OLD LOAN ( $CF = -0.300$ )

# Binomial model: Dynamical hedging/arbitrage

- ❖ We follow a (random) share price  $S$  trajectory: in NEXT STEP ( $n=3$ ) the share price  $\uparrow$



THIS IS THE OPTION  
EXPIRY DATE !!!  
( $N=3$ )

- ❖ We CLOSE ALL OPEN POSITIONS: we SELL  $+0.545$  SHARES @  $S=1.200$  and we REPAY THE LOAN:  $-0.413 \times 1.1 = -0.454$
- ❖ As the option SELLER (SHORT) we MUST PAY the PAYOFF:  $c(3)_2 = 0.200$



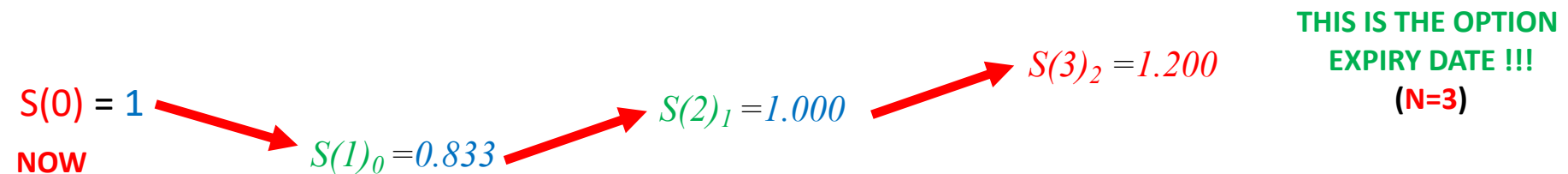
PREVIOUS POSITIONS ( $n=2$ ):  
SHORT call:  $-c$   
LONG shares:  $+0.545 S$   
Loan:  $B=-0.413$   
CURRENT NET WORTH\* = 0!  
\*IF OPTION PRICES ARE FAIR:  $c = \Delta \times S + B$   
\*\*If one closes all positions one gets CF=0

- ❖ NOTE that the current  $CF(3) = 0$  again!, as the INFLOWS from SELLING THE SHARES ( $CF = +0.545 \times 1.200 = +0.655$ ) exactly EQUAL OUTFLOWS FOR REPAYING THE LOAN ( $CF = -0.455$ ) AND THE OPTION'S PAYOFF ( $CF = -0.200$ )

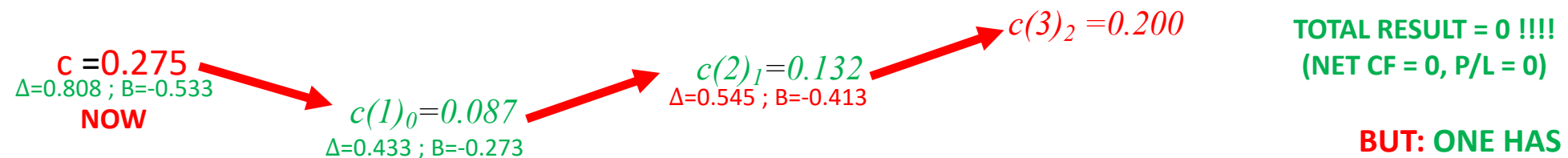


# Binomial model: Dynamical hedging/arbitrage

❖ We have followed a (random) share price  $S$  trajectory



- ❖ In each time STEP the NET CF from trading our TOTAL POSITION (SHORT OPTION + LONG REPLICATING PORTFOLIO) WAS ZERO ! The result is the same for ANY possible trajectory !
- ❖ As the market is DYNAMICALLY COMPLETE, we have successfully managed to construct a DYNAMICALLY ADJUSTED PORTFOLIO of shares and loan which replicated the option perfectly.



- ❖ If the option was (NOW) mispriced, one could make ARBITRAGE PROFITS ! BUT: ONE HAS TO KNOW THE BINOMIAL PROCESS PARAMETERS  $u$  &  $d$  !!!
- if  $c > 0.275$  one sells (NOW) the option at a higher price making DYNAMICAL arbitrage profit i.e. KNOW FUTURE SHARE VOLATILITY !!!
  - if  $c < 0.275$  one inverses all transactions (positions) and also makes DYNAMICAL arbitrage profit

# Binomial model: American options & flexibility

The **Binomial\* model** can be easily adjusted for:

❖ **Various payoff functions** – if payoff depends only on the final share price  $S(N)$  one simply has to adjust option prices at maturity (before iterative backward procedure)

○ E.g. for put options with the exercise price  $X$  one has:  $p(N)_j = \max(X - S u^j d^{N-j} ; 0)$

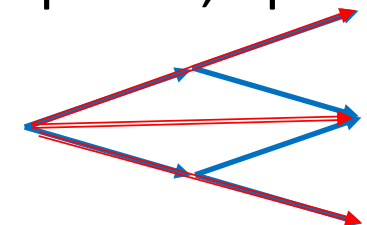
❖ **American options** and early exercise – in each backward Binomial step one has to check the current payoff from the option (i.e. from early exercise)

○ E.g. for American Put options:  $P(n)_j = \max(X - S u^j d^{n-j} ; p(n)_j)$ , where:  $p(n)_j = \frac{g P(n+1)_{j+1} + (1-g)P(n+1)_j}{(1+R)}$  „standard” pricing formula

❖ **Non-constant risk-free rate  $R\%$  and volatility ( $u$  &  $d$ )** – parameters can change in each step of the price evolution

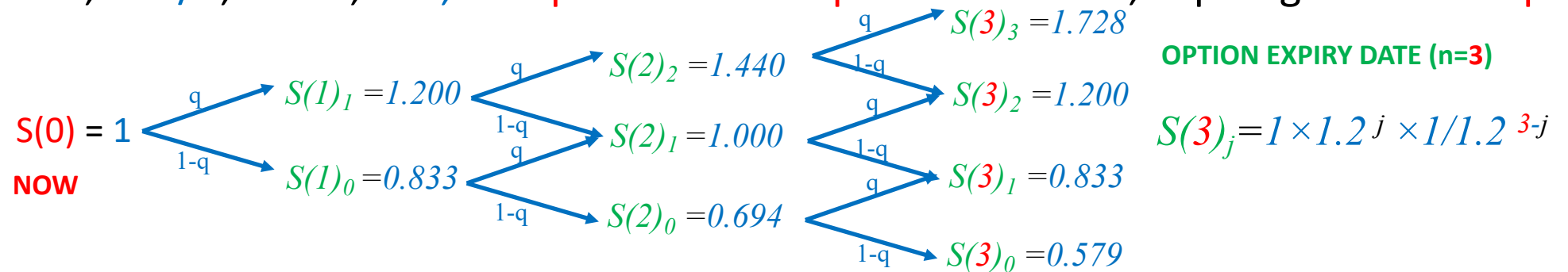
❖ **(Continuous) dividend/other payments** (this works well e.g. for F/X options, options on futures, ...)  $\Rightarrow$  see **Problems: Set 6**

\*One can also construct a (more general) **Trinomial model** (with three-valence graph)  
This way in a single step one can get two Binomial steps  $\Rightarrow$  **better convergence**.



# Binomial model: Example (European binary)

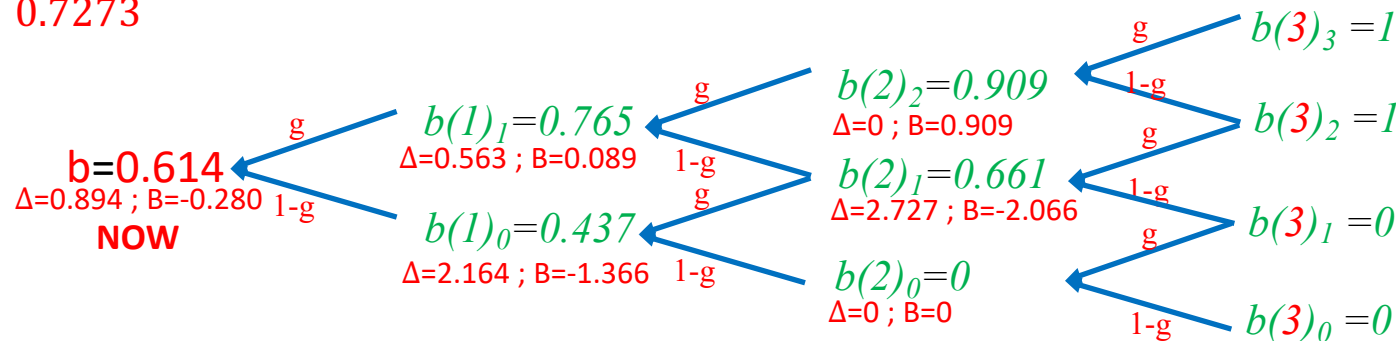
❖ Assume:  $u=1.2$ ,  $d=1/u$ ,  $R=0.1$ ,  $S=1$ , **European BINARY option** with  $X = 1$ , expiring in  $N = 3$  steps



❖ The **BINARY** option pays 1 if  $S(N) > X$  and 0 otherwise:  $b(N)_j = \Theta(S(N)_j - X)$

Heaviside Theta

$$g = \frac{(1 + R) - d}{u - d} = 0.7273$$



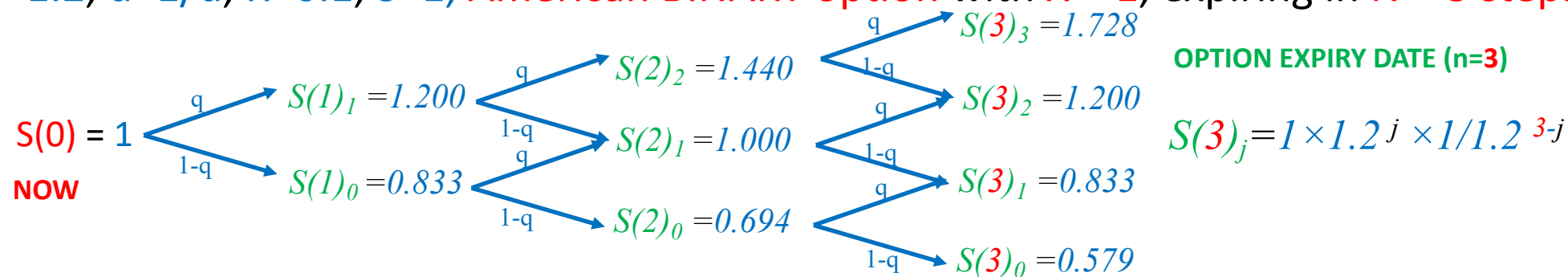
$$b(n)_j = \frac{g b(n+1)_{j+1} + (1-g)b(n+1)_j}{(1+R)}$$

$$\Delta(n)_j = \frac{b(n+1)_{j+1} - b(n+1)_j}{S(n)_j(u-d)}$$

$$B(n)_j = -\frac{d b(n+1)_{j+1} - u b(n+1)_j}{(u-d)(1+R)}$$

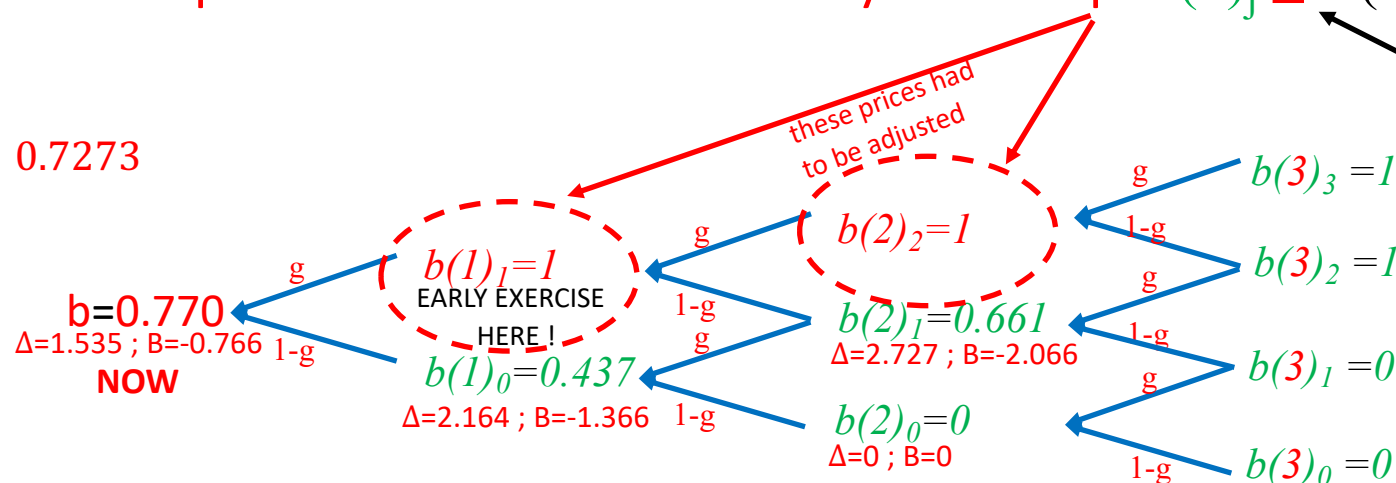
# Binomial model: Example (American binary)

❖ Assume:  $u=1.2$ ,  $d=1/u$ ,  $R=0.1$ ,  $S=1$ , **American BINARY option** with  $X = 1$ , expiring in  $N = 3$  steps



❖ The **American option** can be exercised at any time step:  $b(n)_j \geq \Theta(S(n)_j - X)$  and it is **worth more !**

$$g = \frac{(1 + R) - d}{u - d} = 0.7273$$



$$b(n)_j = \frac{g b(n+1)_{j+1} + (1-g)b(n+1)_j}{(1+R)}$$

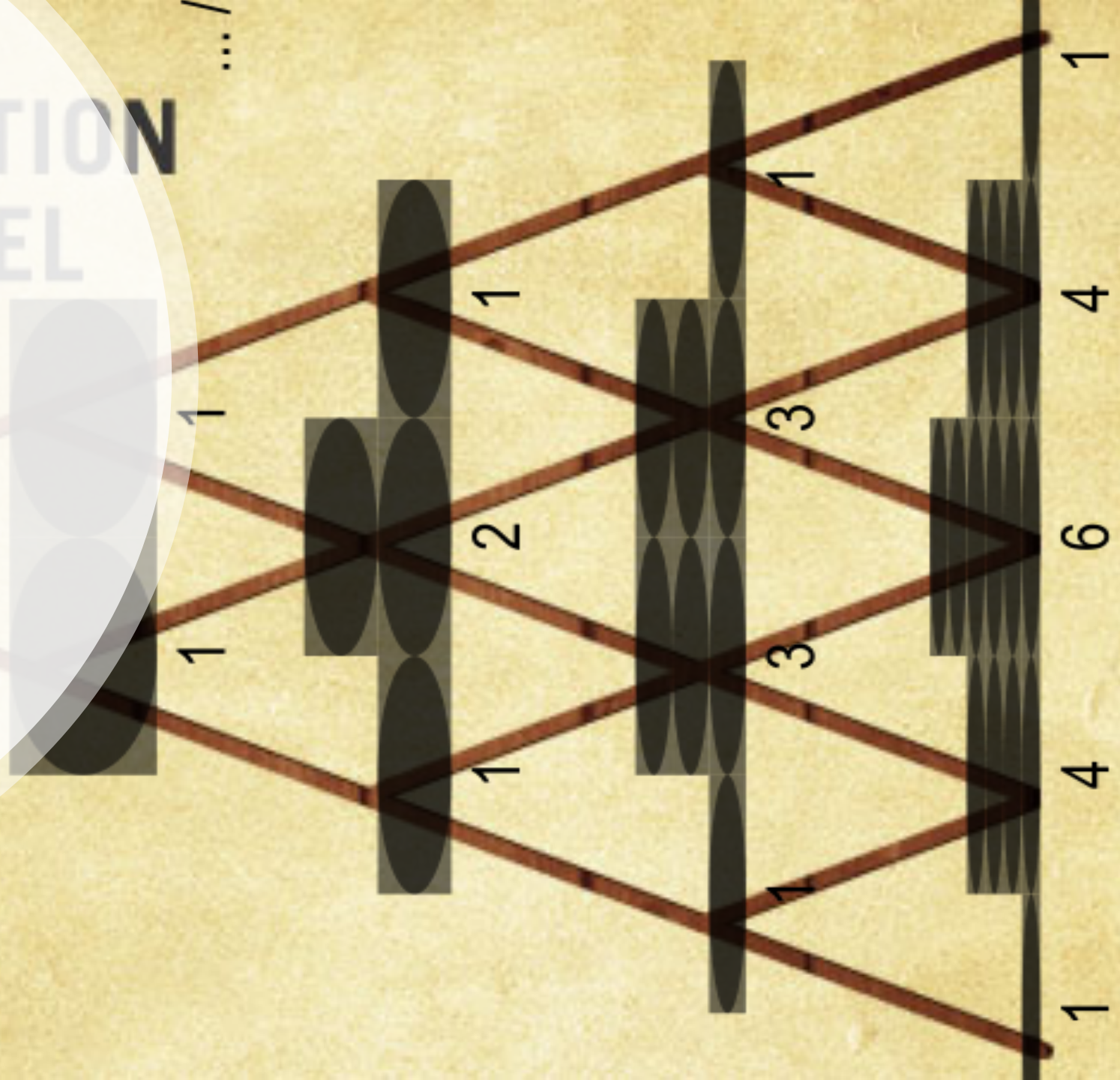
$$\Delta(n)_j = \frac{b(n+1)_{j+1} - b(n+1)_j}{S(n)_j(u-d)}$$

$$B(n)_j = -\frac{d b(n+1)_{j+1} - u b(n+1)_j}{(u-d)(1+R)}$$



# Option pricing in Binomial Model

- ❖ Stochastic processes
- ❖ One-step Binomial Model
- ❖ Multi-step Binomial Model
- ❖ “Analytic” formula for European options
- ❖ Risk-neutral / martingale pricing



# “Analytic” formula for European options

- ❖ We now want to price a (**general**) **European option**:  $V$ , with the payoff function depending only on the final share price  $S(N)$  at expiration (i.e. after  $N$  Binomial time-steps):  $V(N) \equiv V(S(N))$
- ❖ We have shown that, in the  **$N$ -step Binomial model**, for such options one has:

○ At expiration ( $n=N$ ):  $S(N)_j = S u^j d^{N-j} \Rightarrow V(N)_j = V(S u^j d^{N-j})$ ,  $j = 0, \dots, N$

○ For any  $0 \leq n < N$ :  $V(n)_j = \frac{g V(n+1)_{j+1} + (1-g)V(n+1)_j}{(1+R)}$ , where:  $g = \frac{(1+R) - d}{u - d}$

- ❖ It is easy to show\* that the current option price is given by:

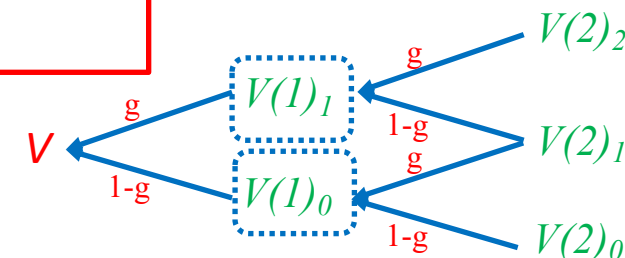
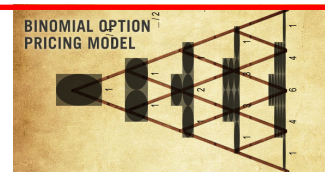
$$V = (1+R)^{-N} \sum_{j=0}^N \binom{N}{j} g^j (1-g)^{N-j} V(S u^j d^{N-j})$$

KNOWN PAYOFF FUNCTION !!!

\*EXAMPLE: for  $N=2$  one has:

$$V = \frac{g \frac{g V(2)_2 + (1-g)V(2)_1}{(1+R)} + (1-g) \frac{g V(2)_1 + (1-g)V(2)_0}{(1+R)}}{(1+R)} = \frac{g^2 V(2)_2 + 2g(1-g)V(2)_1 + (1-g)^2 V(2)_0}{(1+R)^2}$$

How many paths lead to  $j$ -th price



# “Analytic” formula for European options

❖ We now want to price a (**general**) **European option**:  $V$ , with the payoff function depending only on the final share price  $S(N)$  at expiration (i.e. after  $N$  Binomial time-steps):  $V(N) \equiv V(S(N))$

❖ We have shown that, in the **N-step Binomial model**, for such options one has:

○ At expiration ( $n=N$ ):  $S(N)_j = S u^j d^{N-j} \Rightarrow V(N)_j = V(S u^j d^{N-j})$ ,  $j = 0, \dots, N$

○ For any  $0 \leq n < N$ :  $V(n)_j = \frac{g V(n+1)_{j+1} + (1-g)V(n+1)_j}{(1+R)}$ , where:  $g = \frac{(1+R)-d}{u-d}$

KNOWN PAYOFF  
FUNCTION !!!

❖ It is easy to show\* that the current option price is given by:

$$V = (1+R)^{-N} \sum_{j=0}^N \binom{N}{j} g^j (1-g)^{N-j} V(S u^j d^{N-j})$$

❖ This can be used as the (**European**) **option pricing “analytic” formula**, which is simpler to use than going backward the Binomial tree

❖ For the **European binary option** Example (discussed above) one simply has:  $g = \frac{(1+R)-d}{u-d} = 0.7273$

$$V = (1+0.1)^{-3} \sum_j \binom{3}{j} g^j (1-g)^{3-j} \mathbb{1}_{(1 \times 1.2^j \times 1/1.2^{3-j} - 1)} = 1.1^{-3} \left( \binom{3}{3} g^3 \times 1 + \binom{3}{2} g^2 (1-g) \times 1 \right) = 0.614_{44}$$



# “Analytic” formula for European options

$$V = (1 + R)^{-N} \sum_{j=0}^N \binom{N}{j} g^j (1 - g)^{N-j} V(S u^j d^{N-j})$$

❖ In particular, for the **European call** option ( $V \equiv c$ ) one has:

$$c = (1 + R)^{-N} \sum_{j=0}^N \binom{N}{j} g^j (1 - g)^{N-j} \max(S u^j d^{N-j} - X; 0) = \dots$$

$$\dots = (1 + R)^{-N} \sum_{j=a}^N \binom{N}{j} g^j (1 - g)^{N-j} (S u^j d^{N-j} - X) = \dots$$

$a$ : the smallest integer such that:  $S u^a d^{N-a} > X \Rightarrow a = \lceil \ln(X/(S d^N)) / \ln(u/d) \rceil$

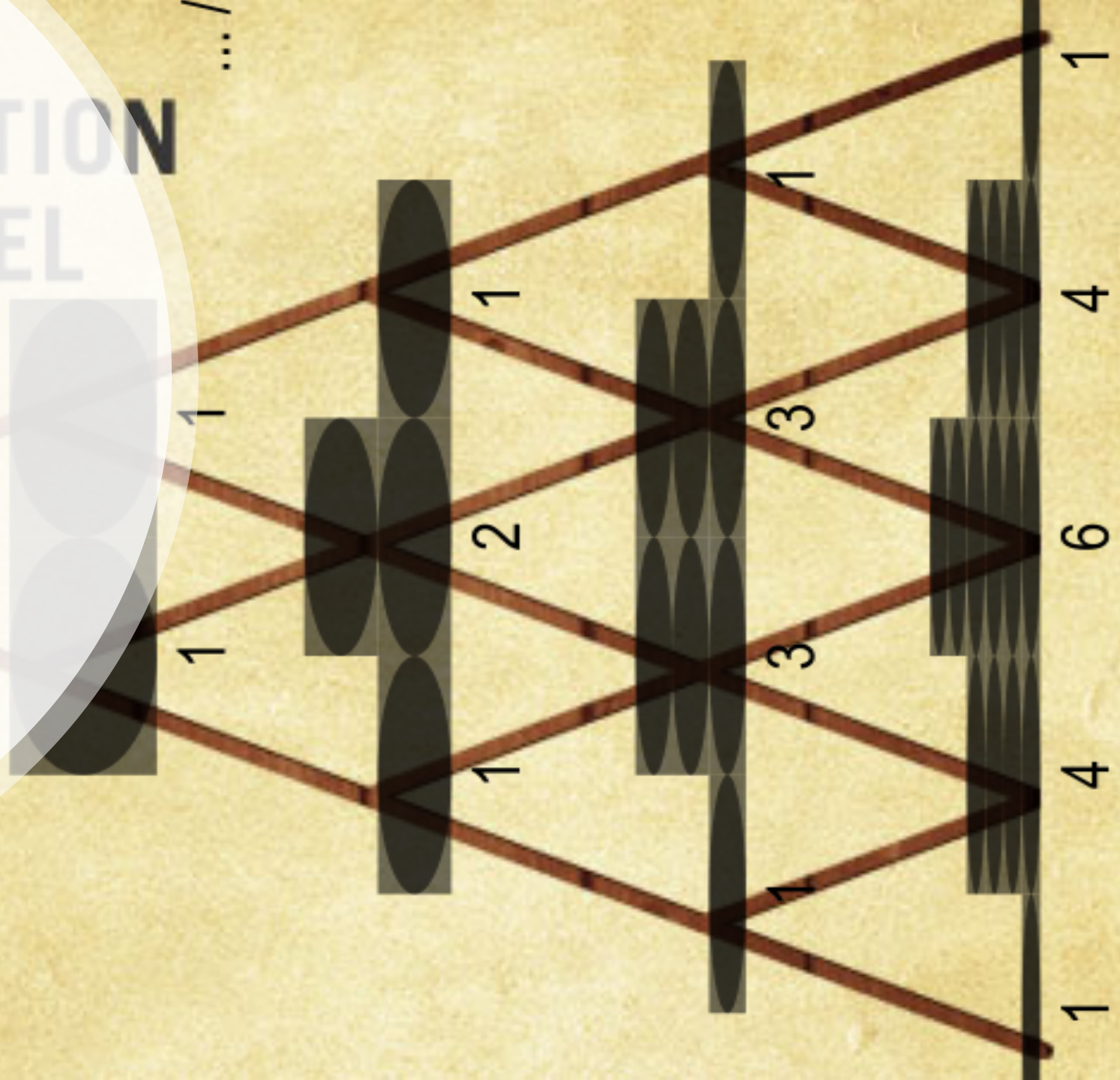
$$\dots = S \sum_{j=a}^N \binom{N}{j} \underbrace{\left(\frac{ug}{1+R}\right)^j}_{g'} \underbrace{\left(\frac{d(1-g)}{1+R}\right)^{N-j}}_{1-g'} - X (1 + R)^{-N} \sum_{j=a}^N \binom{N}{j} g^j (1 - g)^{N-j}$$

As  $g$  (and  $g'$ ) can be interpreted as **probability** (see next section)  
 $\Phi(N; a; g) = 1 - \text{CDF}(\text{Binom.}(N, g))$

$$c = S \Phi(N; a; g') - X (1 + R)^{-N} \Phi(N; a; g)$$

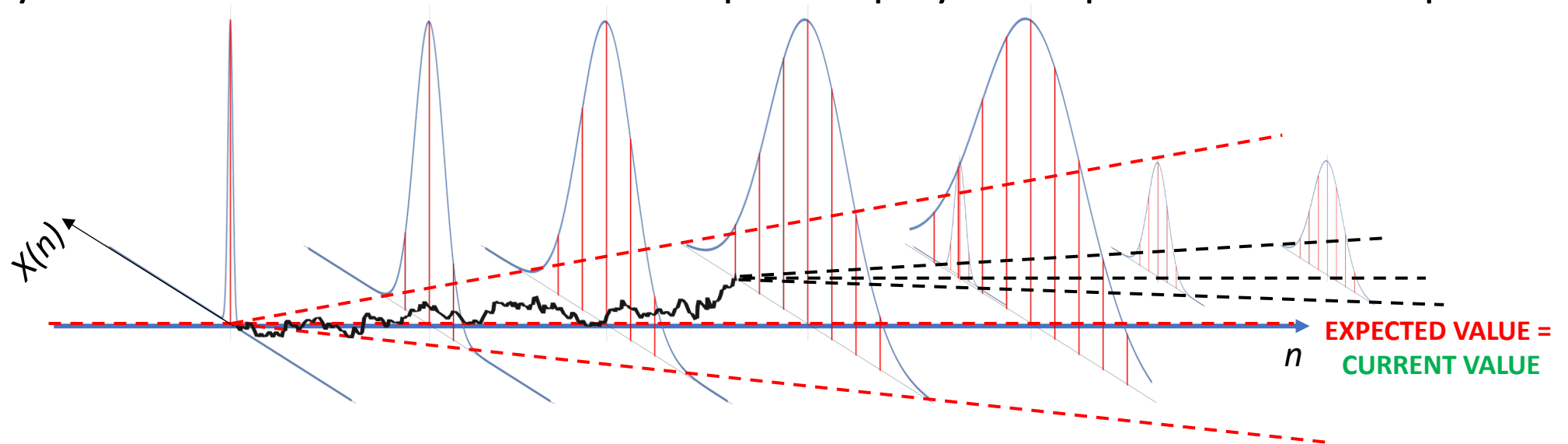
# Option pricing in Binomial Model

- ❖ Stochastic processes
- ❖ One-step Binomial Model
- ❖ Multi-step Binomial Model
- ❖ “Analytic” formula for European options
- ❖ Risk-neutral / martingale pricing



# Martingales (short introduction)

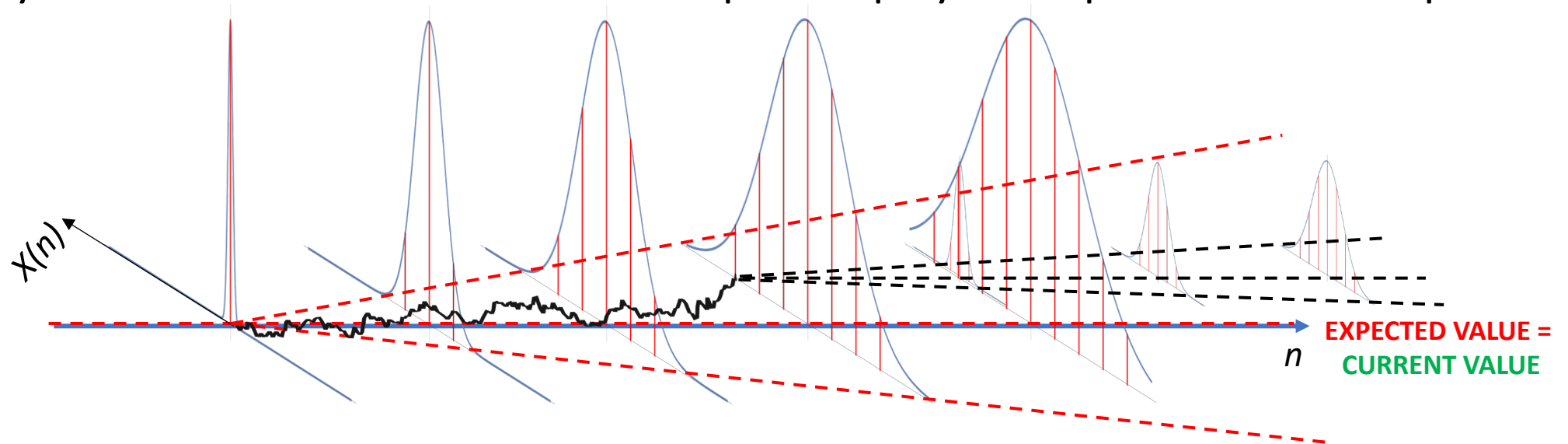
- ❖ **Martingales** are stochastic processes representing “fair games”, i.e. in each step the probability to win or loose is such that the expected player’s capital = current capital



- ❖ One can thus talk about **conditional expectation** of the **next value** (observation)  $X(n+1)$  in a stochastic process sequence, given all **prior values** (observations):  $X(0), X(1), \dots, X(n)$
- ❖ For a **martingale** one has:  $E(X(n+1) \mid X(0), X(1), \dots, X(n)) = X(n)$  for any step  $n$
- ❖ For a **continuous-time martingale** it is replaced by:  $E(X(t') \mid \{X(\tau), \tau \leq t\}) = X(t): \forall t \leq t'$

# Martingales (short introduction)

- ❖ **Martingales** are stochastic processes representing “fair games”, i.e. in each step the probability to win or loose is such that the expected player’s capital = current capital



- ❖ A stochastic process  $\{X(n)\}$  can be also a **martingale wrt another process  $\{Y(n)\}$**  if:  

$$E(X(n+1) \mid Y(0), Y(1), \dots, Y(n)) = X(n)$$
for any step  $n$
- ❖ All information available at a time  $n$ , i.e.  $Y(0), Y(1), \dots, Y(n)$  is often called a “**filtration**”
- ❖ A process is “**adapted to filtration**” if for every realization:  $X(n)$  is known at time  $n$

# Risk-neutral / Martingale pricing

$$\sum_t \langle PV(CF(t)) \rangle = 0 \quad (4)$$

- ❖ In **Lecture 5** we showed that one can use Eqn. (4) to price some derivative instruments (e.g. forwards) if one adjusts the stochastic process of the underlying asset (e.g. the share) price in such a way that its present value is a MARTINGALE, i.e.:  $\langle PV(S(T)) \rangle = S(0)$  for each time T
- ❖ In this Lecture we have shown that in the **one-step Binomial model** one has:

$$c = \frac{g c(1)_u + (1 - g) c(1)_d}{(1 + R)} \quad (*), \text{ where: } g = \frac{(1 + R) - d}{u - d}$$

- ❖ NOTE, that due to (static) **ARBITRAGE arguments**:  $d \leq 1 + R \leq u$ 
  - If  $d > 1 + R$ : one can take a LOAN for S @ R% and BUY 1 SHARE @ S (NET CF=0), after one step one SELLS the share getting (at least) Sd and REPAYS the LOAN paying S(1+R) (NET CF > 0)  $\Rightarrow$  **ARBITRAGE**
  - If  $u < 1 + R$ : one SELLS 1 SHARE SHORT @ S and makes DEPOSIT for S @ R% (NET CF=0), after one step one BUYS the share back paying (at most) Su and from the DEPOSIT one gets S(1+R) (NET CF > 0)  $\Rightarrow$  **ARBITRAGE**
- ❖ Thus:  $0 \leq g \leq 1$  is a **PROBABILITY**, and Eqn. (\*) becomes:  $c = \langle PV(c(1)) \rangle_g$ , so it looks exactly like Eqn. (4)



# Risk-neutral / Martingale pricing

$$\sum_t \langle PV(CF(t)) \rangle = 0 \quad (4)$$

- ❖ In **Lecture 5** we showed that one can use Eqn. (4) to price some derivative instruments (e.g. forwards) if one adjusts the stochastic process of the underlying asset (e.g. the share) price in such a way that its present value is a MARTINGALE, i.e.:  $\langle PV(S(T)) \rangle = S(0)$  for each time  $T$
- ❖ The same applies for the **N-step Binomial model** where one has:

$$c(n)_j = \frac{\overbrace{g c(n+1)_{j+1}}^{c_u} + (1-g) \overbrace{c(n+1)_j}^{c_d}}{(1+R)}, \quad g = \frac{(1+R) - d}{u - d}$$

KNOWN CF(N) AT EXPIRY !  
 $c(N)_j = \max(S u^j d^{N-j} - X; 0)$

❖ Resulting in (see p. 43):  $c = \sum_{j=0}^N \underbrace{\binom{N}{j} g^j (1-g)^{N-j}}_{\tilde{P}_g(N)_j} \underbrace{(1+R)^{-N} c(N)_j}_{PV(CF(N))} \quad (**)$

For  $0 \leq g \leq 1$  this is a **probability**  
 in **Binomial(N;g)** distribution

- ❖ So again Eqn. (\*\*) becomes:  $c = \langle PV(c(N)) \rangle_g$  in accordance with Eqn. (4)

# Risk-neutral / Martingale pricing

$$\sum_t \langle PV(CF(t)) \rangle = 0 \quad (4)$$

- ❖ In **Lecture 5** we showed that one can use Eqn. (4) to price some derivative instruments (e.g. forwards) if one **adjusts the stochastic process of the underlying asset** (e.g. the **share**) price in such a way that its **present value** is a **MARTINGALE**, i.e.:  $\langle PV(S(T)) \rangle = S(0)$  for each time T
- ❖ If, in the **Binomial** model, the (true) share price evolution **probability**  $q$  is replaced by a (fictitious) **probability**  $g = \frac{(1+R)-d}{u-d}$ , then one can explicitly check that:

$$\langle PV(S(n+1)) \rangle_g = g \frac{S(n) u}{(1+R)^{n+1}} + (1-g) \frac{S(n) d}{(1+R)^{n+1}} = \frac{S(n)}{(1+R)^n} = PV(S(n))$$

❖ So:  $E(PV(S(n+1)) \mid S(0), PV(S(1)), \dots, PV(S(n))) = PV(S(n))$

and the **(discounted) share price** process becomes a **MARTINGALE\*** !

\*This also automatically implies:  $\langle PV(S(n)) \rangle_g = \sum_j \tilde{P}_g(n)_j \frac{S(n)_j}{(1+R)^n} = S(0)$  for any  $n$ , where:  $\tilde{P}_g(n)_j = \binom{n}{j} g^j (1-g)^{n-j}$  so if the share price evolution is driven by **martingale probability**  $g \Rightarrow$  **expected rate of return from shares = risk-free rate**

**“RISK-NEUTRAL” PRICING !!!**



# Risk-neutral / Martingale pricing

$$\sum_t \langle PV(CF(t)) \rangle = 0 \quad (4)$$

- ❖ In **Lecture 5** we showed that one can use Eqn. (4) to price some derivative instruments (e.g. forwards) if one **adjusts the stochastic process of the underlying asset** (e.g. the **share**) price in such a way that its **present value** is a **MARTINGALE**, i.e.:  $\langle PV(S(T)) \rangle = S(0)$  for each time T
- ❖ The option is a **derivative instrument**, whose price (stochastic process) depends on the share price (stochastic process)

- if we artificially adjust the (discounted) **share price** process  $\{PV(S(t))\}_t$  by changing **probability q** → **probability g** such that it becomes a **martingale** and thus Eqn. (4) is fulfilled by S(t), or in general:

$$E(PV(S(t')) \mid \{PV(S(\tau)), \tau \leq t\}) = PV(S(t)): \forall t \leq t'$$

- then also the **option price** process  $\{c(t)\}_t$  becomes a **martingale wrt (filtration of) the share price** process, and thus Eqn. (4) is fulfilled also by c(t), or in general:

$$E(PV(c(t')) \mid \{PV(S(\tau)), \tau \leq t\}) = PV(c(t)): \forall t \leq t'$$

- this is **TRUE ONLY IF** the derivative (e.g. option) **MARKET IS (DYNAMICALLY) COMPLETE**, i.e. if options can be replicated by a portfolio of shares and risk-free instruments (loans/deposits)

- ❖ Summing up\* (in the Binomial model):  $\langle PV(S(N)) \rangle_g = S(0) \Rightarrow c(0) = \langle PV(c(S(N))) \rangle_g$

\*The **expectation value**  $\langle \rangle_g$  is computed with probability measure  $\tilde{P}_g(n)_j = \binom{n}{j} g^j (1-g)^{n-j}$ , where:  $g = \frac{(1+R)-d}{u-d}$

$$\sum_t \langle PV(CF(t)) \rangle = 0 \quad (4)$$

# Radon-Nikodým derivative

- ❖ We have discussed that if the market is (dynamically) complete then options (and in general derivative instruments) can be priced using Eqn. (4) but one has to **change** the **real probability measure**  $\mathcal{P}_q$  into the **martingale probability measure**  $\tilde{\mathcal{P}}_g$  when computing  $\langle \quad \rangle$ .
- ❖ In order to do so one can use the, so called, **(discrete) Radon-Nikodým derivative** of  $\tilde{\mathcal{P}}_g$  wrt  $\mathcal{P}_q$ :

$$Z \equiv \frac{\tilde{\mathcal{P}}_g}{\mathcal{P}_q}$$

- ❖ For a continuous prob. distrib. one defines a **(continuous) Radon-Nikodým derivative**:  $Z \equiv \frac{d\tilde{\mathcal{P}}_g}{d\mathcal{P}_q}$
- ❖  $Z$  is a **random variable** under probability measure  $\mathcal{P}_q$  with the following **properties**:
  - $\Pr(Z > 0) = 1$
  - $\langle Z \rangle_q = 1$
  - For any random variable  $V$ :  $\langle V \rangle_g = \langle V Z \rangle_q \iff$  This enables to compute the expectation value in (4)\* using prob.  $\mathcal{P}_q$  instead of  $\tilde{\mathcal{P}}_g$  if one knows  $Z$

\*E.g. for a continuous prob. distrib. one has:  $\langle V \rangle_g = \int V d\tilde{\mathcal{P}}_g = \int V \frac{d\tilde{\mathcal{P}}_g}{d\mathcal{P}_q} d\mathcal{P}_q = \int V Z d\mathcal{P}_q = \langle V Z \rangle_q$

# Radon-Nikodým derivative

$$\langle V \rangle_g = \langle V Z \rangle_q$$

$$\sum_t \langle PV(CF(t)) \rangle = 0 \quad (4)$$

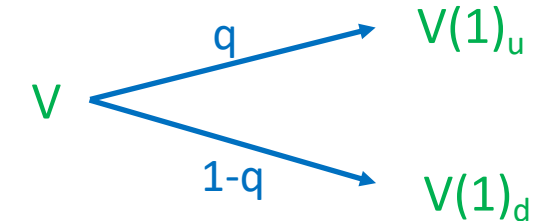
- ❖ We have discussed that if the market is (dynamically) complete then options (and in general derivative instruments) can be priced using Eqn. (4) but one has to **change** the **real probability measure**  $\mathcal{P}_q$  into the **martingale probability measure**  $\tilde{\mathcal{P}}_g$  when computing  $\langle \quad \rangle$ .
- ❖ In order to do so one can use the, so called, **(discrete) Radon-Nikodým derivative** of  $\tilde{\mathcal{P}}_g$  wrt  $\mathcal{P}_q$ :

$$Z \equiv \frac{\tilde{\mathcal{P}}_g}{\mathcal{P}_q}$$

- ❖ For a (discrete) **one-step Binomial** process one simply has:

- $Z(1)_u = \frac{g}{q}$  with probability  $q$  and  $Z(1)_d = \frac{(1-g)}{(1-q)}$  with probability  $1-q$

- So trivially:  $\langle V Z \rangle_q = V(1)_u Z(1)_u q + V(1)_d Z(1)_d (1-q) = V(1)_u g + V(1)_d (1-g) = \langle V \rangle_g$



- ❖ For a **multi-step Binomial** process for each  $n$ :

- $Z(n)_j = \frac{\tilde{\mathcal{P}}_g(n)_j}{\mathcal{P}_q(n)_j} = \left(\frac{g}{q}\right)^j \left(\frac{1-g}{1-q}\right)^{n-j}$  with probability  $\mathcal{P}_q(n)_j$ ,  $j = 0, \dots, n$

- One can show that the **Radon-Nikodým derivative** process  $\{Z(n)\}_n$  is a **martingale** under measure  $\mathcal{P}_q$   
 $E(Z(n+1) \mid Z(0), \dots, Z(n)_j) = Z(n)_j$

# Summary

