

Multi-Period Asset Pricing

Part 4

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Arbitrage-free pricing in multi-period binomial model

Why multi-period models?

Backward induction

Why multi-period models?

Single period binomial model: complete, but unrealistic.

Single period trinomial model: more realistic, but incomplete.

Indeed, let d , h , and u be the relative changes in the stock price and r be the interest rate such that

$$d < 1 + r < u, \quad d < h < u.$$

For risk-neutral probabilities \tilde{p}_i , $i = 1, 2, 3$, we get the system:

$$\tilde{p}_i > 0, \quad i = 1, 2, 3, \tag{a}$$

$$\tilde{p}_1 + \tilde{p}_2 + \tilde{p}_3 = 1, \tag{b}$$

$$\tilde{p}_1 d + \tilde{p}_2 h + \tilde{p}_3 u = 1 + r \quad \left(\tilde{\mathbb{E}}(S_1) = S_0(1 + r) \right) \tag{c}$$

Why multi-period models?

From (b) + (c) we deduce that

$$\begin{aligned}\tilde{p}_2 &= \frac{u - (1 + r) - (u - d)\tilde{p}_1}{u - h}, \\ \tilde{p}_3 &= \frac{1 + r - h + (h - d)\tilde{p}_1}{u - h}.\end{aligned}$$

Accounting for (a) we obtain that

$$\max\left(0, \frac{h - (1 + r)}{h - d}\right) < \tilde{p}_1 < \frac{u - (1 + r)}{u - d}.$$

Thus, there are ∞ -many solutions and

FTAPs \implies NA + incomplete.

Why multi-period models?

There are 2 ways to get more realistic and still complete model:

1. Add extra traded securities; for instance, liquid options.
2. Trade dynamically, that is, use multi-period models.

Remark (Time scales of financial models)

Single-period: easy to understand and use, but not realistic.

Multi-period: elementary to learn, very realistic, but not convenient for computations.

Continuous-time: quite realistic, very convenient for computations, but difficult to master (Ito's Calculus!).

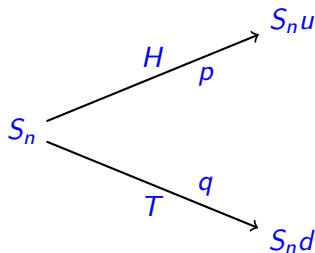
Multi-period binomial model

There are $N + 1$ times: $0, 1, \dots, N$.

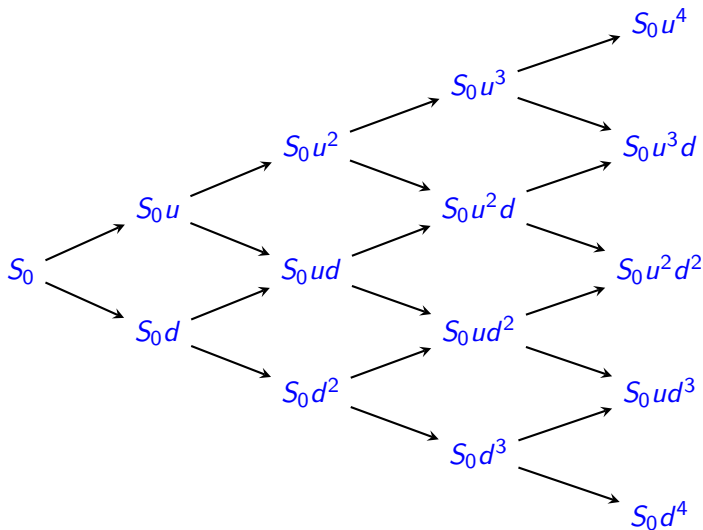
Bank account with interest rate $r > -1$:

$$X_n \text{ at } t = n \longrightarrow X_n(1 + r) \text{ at } t = n + 1.$$

Stock with parameters u (“up”) and d (“down”) such that $u > d > 0$:



Multi-period binomial model



Probability space for binomial model

Probability space: (Ω, \mathbb{P}) , where elementary event $\omega \in \Omega$ is a *trajectory*:

$$\omega = (\omega_1, \dots, \omega_N), \quad \omega_i \in \{T, H\}.$$

The number of elementary events:

$$|\Omega| = 2^N. \quad (\text{Grows very fast!})$$

As always, we assume that \mathbb{P} is strictly positive:

$$\mathbb{P}(\omega) > 0, \quad \omega \in \Omega.$$

Of course, we also have that

$$\sum_{\omega \in \Omega} \mathbb{P}(\omega) = 1.$$

Adapted sequences in binomial model

Definition

A sequence of random variables $(X_n)_{0 \leq n \leq N}$ is **adapted** if for every $t = n$,

$$X_n = X_n(\omega_1, \dots, \omega_n),$$

that is, X_n depends only of the first n letters $(\omega_1, \dots, \omega_n)$ in $\omega = (\omega_1, \dots, \omega_N)$.

Examples

The following sequences are adapted: (S_n) , $(\max_{k \leq n} S_k)$,

(X_n) : the capital of a strategy,

(Δ_n) : the number of stocks in a strategy.

Arbitrage in binomial model

Lemma

In the multi-period binomial model,

$$NA \iff d < 1 + r < u.$$

Proof.

For a multi-period model, an arbitrage exists *globally* if and only if it exists *locally* in one of the branches. The result now follows from the single period case. \square

European options in binomial model

A European option with maturity N is determined by its payoff

$$V_N = V_N(\omega) = V_N(\omega_1, \dots, \omega_N).$$

Examples

Lookback option: $V_N(\omega) = M_N - S_N(\omega)$, where

$$M_N = \max_{0 \leq n \leq N} S_n(\omega).$$

Asian call option: $V_N(\omega) = \max(A_N - K, 0)$, where

$$A_N = \frac{1}{N+1} \sum_{n=0}^N S_n.$$

AFP in binomial model

!AFP = Replication

Replicating strategy:

$$X_0 \xrightarrow{(\Delta_n)} \underbrace{X_N = V_N}_{\text{known}}$$

or, more precisely,

$$X_0 \xrightarrow{\Delta_0} X_1 \xrightarrow{\Delta_1} X_2 \xrightarrow{\Delta_2} \cdots X_{N-1} \xrightarrow{\Delta_{N-1}} \underbrace{X_N = V_N}_{(\text{known})}$$

To compute it, we move **BACKWARD** in time:

$$\underbrace{X_0 \leftarrow \cdots X_n}_{\text{unknown}} \leftarrow \underbrace{X_{n+1} \leftarrow \cdots \overbrace{X_N = V_N}^{\text{boundary condition}}}_{\text{known}}$$

AFP in binomial model

One-step iteration:

$$\underbrace{X_n}_{?} \xrightarrow{\Delta_n - ?} \underbrace{X_{n+1}}_{\text{known}}$$

Given the history $(\omega_1, \dots, \omega_n)$, we obtain that

$$X_{n+1}(\omega_{n+1} = H) = (X_n - \Delta_n S_n)(1 + r) + \Delta_n S_n u,$$

$$X_{n+1}(\omega_{n+1} = T) = (X_n - \Delta_n S_n)(1 + r) + \Delta_n S_n d,$$

which is the same system as in the single period case.

AFP in binomial model

We deduce that

$$\Delta_n = \frac{X_{n+1}(\omega_{n+1} = H) - X_{n+1}(\omega_{n+1} = T)}{S_{n+1}(\omega_{n+1} = H) - S_{n+1}(\omega_{n+1} = T)},$$
$$X_n = \frac{1}{1+r} (\tilde{p}X_{n+1}(\omega_{n+1} = H) + \tilde{q}X_{n+1}(\omega_{n+1} = T)),$$

where as in the single period case,

$$\tilde{p} = \frac{1+r-d}{u-d}, \quad \tilde{q} = 1 - \tilde{p} = \frac{u-(1+r)}{u-d}.$$

AFP in binomial model

These formulas together with *boundary condition*:

$$X_N(\omega) = V_N(\omega), \quad \omega \in \Omega,$$

form the algorithm of **backward induction** for the AFP in the N -period binomial model.

Remark

As the algorithm shows, we can find a replicating strategy for *every* European option. Hence, the N -period binomial model is *complete*.

AFP in binomial model

Remark

The equation for the backward induction can be written in a more compact way:

$$X_n = \mathcal{R}_n(X_{n+1}),$$

where the operator \mathcal{R}_n acts on $X_{n+1} = X_{n+1}(\omega_1, \dots, \omega_n, \omega_{n+1})$ as follows:

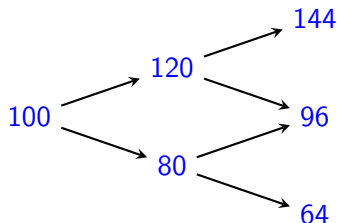
$$\begin{aligned} \mathcal{R}_n(X_{n+1})(\omega_1, \dots, \omega_n) = \\ \frac{1}{1+r} (\tilde{p}X_{n+1}(\omega_1, \dots, \omega_n, H) + \tilde{q}X_{n+1}(\omega_1, \dots, \omega_n, T)). \end{aligned}$$

The operator \mathcal{R}_n plays a crucial role in software libraries for pricing derivatives. It is called the *rollback operator*.

Pricing of standard call

Problem

The bank pays the interest rate $r = 0.1$. The stock price follows the two-period binomial model with $S_0 = 100$, $u = 1.2$, $d = 0.8$:



For the call option with maturity $N = 2$ and strike $K = 100$, compute at $t = 0$

1. the arbitrage-free price V_0 ,
2. the number of stocks Δ_0 in the replicating strategy.

Pricing of standard call

Solution

The one-step risk-neutral probabilities are given by

$$\tilde{p} = \frac{1 + r - d}{u - d} = \frac{3}{4}, \quad \tilde{q} = \frac{u - 1 - r}{u - d} = \frac{1}{4}.$$

Backward induction:

Time 2: The price of the stock and the value of the option are given in the following table:

ω	stock	call
HH	144	44
TH	96	0
HT	96	0
TT	64	0

Pricing of standard call

Time 1: The value of the call option is given by

$$V_1(H) = \frac{1}{1+r} (\tilde{p}V_2(HH) + \tilde{q}V_2(HT)) = 30,$$

$$V_1(T) = \frac{1}{1+r} (\tilde{p}V_2(TH) + \tilde{q}V_2(TT)) = 0.$$

Time 0: Finally, we obtain that

$$V_0 = \frac{1}{1+r} (\tilde{p}V_1(H) + \tilde{q}V_1(T)) = 20.45,$$

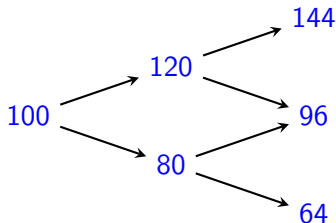
$$\Delta_0 = \frac{V_1(H) - V_1(T)}{S_1(H) - S_1(T)} = \frac{3}{4}.$$

□

Pricing of Asian call

Problem

The bank pays the interest rate $r = 0.1$. The stock price follows the two-period binomial model with $S_0 = 100$, $u = 1.2$, $d = 0.8$:



Compute the AFP of the Asian call with strike $K = 100$ and the payoff

$$V_N = \max \left(\frac{1}{N+1} \sum_{k=0}^N S_k - K, 0 \right).$$

Pricing of Asian call

Solution

The one-step risk-neutral probabilities are given by

$$\tilde{p} = \frac{1 + r - d}{u - d} = \frac{3}{4}, \quad \tilde{q} = 1 - \tilde{p} = \frac{1}{4}.$$

Time 2: The price of the option is given by

$$V_2(HH) = \max \left(\frac{100 + 120 + 144}{3} - 100, 0 \right) = 21.3333,$$

$$V_2(HT) = \max \left(\frac{100 + 120 + 96}{3} - 100, 0 \right) = 5.3333,$$

$$V_2(TH) = \max \left(\frac{100 + 80 + 96}{3} - 100, 0 \right) = 0,$$

$$V_2(TT) = \max \left(\frac{100 + 80 + 64}{3} - 100, 0 \right) = 0.$$

Pricing of Asian call

Time 1: The value of the Asian call option is given by

$$V_1(H) = \frac{1}{1+r} (\tilde{p}V_2(HH) + \tilde{q}V_2(HT)) = 15.7576,$$

$$V_1(T) = \frac{1}{1+r} (\tilde{p}V_2(TH) + \tilde{q}V_2(TT)) = 0.$$

Time 0: The value of the Asian call option is given by

$$V_0 = \frac{1}{1+r} (\tilde{p}V_1(H) + \tilde{q}V_1(T)) = 10.74.$$

□

Arrow-Debreu security

Problem

Let $\hat{\omega} = (\hat{\omega}_1, \dots, \hat{\omega}_N)$ be an outcome. Compute the AFP of the Arrow-Debreu security paying

$$V_N(\omega) = 1_{\{\omega = \hat{\omega}\}} \quad \text{at } t = N.$$

Solution

Let $V_n = V_n(\omega_1, \dots, \omega_n)$ be the AFP at $t = n$.

If $(\omega_1, \dots, \omega_n) \neq (\hat{\omega}_1, \dots, \hat{\omega}_n)$, then

$$V_n = 0.$$

If $(\omega_1, \dots, \omega_n) = (\hat{\omega}_1, \dots, \hat{\omega}_n)$, then

Arrow-Debreu security

$$\begin{aligned} V_n &= \frac{1}{1+r} (\tilde{p} V_{n+1}(\omega_{n+1} = H) + \tilde{q} V_{n+1}(\omega_{n+1} = T)) \\ &= \frac{1}{1+r} (\tilde{p} 1_{\{\hat{\omega}_{n+1}=H\}} + \tilde{q} 1_{\{\hat{\omega}_{n+1}=T\}}) \\ &\quad \times V_{n+1}(\hat{\omega}_1, \dots, \hat{\omega}_n, \hat{\omega}_{n+1}). \end{aligned}$$

We deduce by induction that

$$\begin{aligned} V_0 &= \frac{1}{(1+r)^N} \prod_{n=1}^N (\tilde{p} 1_{\{\hat{\omega}_{n+1}=H\}} + \tilde{q} 1_{\{\hat{\omega}_{n+1}=T\}}) \\ &= \frac{1}{(1+r)^N} \tilde{p}^{\#H(\hat{\omega})} \tilde{q}^{\#T(\hat{\omega})}. \end{aligned}$$

Here $\#H(\omega)$ is the number of letters H and $\#T(\omega)$ is the number of letters T in $\omega = (\omega_1, \dots, \omega_N)$. □

Risk-neutral valuation

The expression for the AFP of an Arrow-Debreu security yields an “explicit” formula for the AFP of any option. Indeed, as

$$V_N(\omega) = \sum_{\hat{\omega} \in \Omega} V_N(\hat{\omega}) 1_{\{\omega = \hat{\omega}\}},$$

we obtain that

$$\begin{aligned} V_0 &= \frac{1}{(1+r)^N} \sum_{\hat{\omega} \in \Omega} V_N(\hat{\omega}) \tilde{p}^{\#H(\hat{\omega})} \tilde{q}^{\#T(\hat{\omega})} \\ &= \frac{1}{(1+r)^N} \sum_{\omega \in \Omega} V_N(\omega) \tilde{p}^{\#H(\omega)} \tilde{q}^{\#T(\omega)}. \end{aligned}$$

Risk-neutral valuation

We observe now that the function

$$\tilde{\mathbb{P}}(\omega) \triangleq \prod_{n=1}^N (\tilde{p}1_{\{\omega_n=H\}} + \tilde{q}1_{\{\omega_n=T\}}) = \tilde{p}^{\#H(\omega)} \tilde{q}^{\#T(\omega)}, \quad \omega \in \Omega,$$

is a strictly positive probability measure. Indeed,

$$\tilde{\mathbb{P}}(\omega) > 0, \quad \omega \in \Omega,$$

$$\sum_{\omega \in \Omega} \tilde{\mathbb{P}}(\omega) = \sum_{i=1, \dots, N} \sum_{\omega_i \in \{H, T\}} \prod_{n=1}^N (\tilde{p}1_{\{\omega_n=H\}} + \tilde{q}1_{\{\omega_n=T\}}) = 1.$$

In terms of $\tilde{\mathbb{P}}$, the expression for V_0 becomes the RNV:

$$V_0 = \frac{1}{(1+r)^N} \sum_{\omega \in \Omega} V_N(\omega) \tilde{\mathbb{P}}(\omega) = \frac{1}{(1+r)^N} \tilde{\mathbb{E}}(V_N).$$