

# MH4514 Financial Mathematics

- Midterm Revision -

Naoki Honda

March 2019

## 1 (10.6) Properties of the Conditional Expectation

### Properties of CE

- If  $G$  depends only on the information contained in  $\mathcal{G}$ ,

$$\mathbb{E}[FG|\mathcal{G}] = G\mathbb{E}[F|\mathcal{G}]$$

- If  $G$  depends only on the information contained in  $\mathcal{G}$ ,

$$\mathbb{E}[G|\mathcal{G}] = G$$

- **Tower Property**

$$\mathbb{E}[\mathbb{E}[F|\mathcal{G}]|\mathcal{H}] = \mathbb{E}[F|\mathcal{H}] \quad (\text{if } \mathcal{H} \subset \mathcal{G})$$

- When  $F$  "does not depend" on the information contained in  $\mathcal{G}$  (more stated, when the random variable  $F$  is *independent* of the  $\sigma$ -algebra  $\mathcal{G}$ ),

$$\mathbb{E}[F|\mathcal{G}] = \mathbb{E}[F]$$

- If  $G$  depends only on  $\mathcal{G}$  and  $F$  is independent of  $\mathcal{G}$ , then

$$\mathbb{E}[h(F, G)|\mathcal{G}] = \mathbb{E}[h(F, x)]_{x=G}$$

## 2 (1) Discrete-Time Martingales

### 2.1 Definition

#### Definition 1.1

An integrable<sup>a</sup>, discrete-time process  $(Z_n)_{n \in \mathbb{N}}$  is a martingale with respect to  $(\mathcal{F}_n)_{n \in \mathbb{N}}$  if  $(Z_n)_{n \in \mathbb{N}}$  is  $\mathcal{F}_n$ -adapted and satisfies the property

$$\mathbb{E}[Z_{n+1} | \mathcal{F}_n] = Z_n, \quad n \in \mathbb{N}$$

---

<sup>a</sup>Integrable means  $\mathbb{E}[|Z_n|] < \infty$  for all  $n \in \mathbb{N}$

### 2.2 Properties

A particular property of martingales is that their expectation is constant over time.

#### Proposition 1.2

Let  $(Z_n)_{n \in \mathbb{N}}$  be martingale, we have

$$\mathbb{E}[Z_n] = \mathbb{E}[Z_0], \quad n \in \mathbb{N}$$

The following Theorem 1.6 is called the Doob's stopping time theorem.

#### Theorem 1.6

Assume that  $(M_n)_{n \in \mathbb{N}}$  is a martingale with respect to  $(\mathcal{F}_n)_{n \in \mathbb{N}}$ . Then the *stopped process*  $(M_{\tau \wedge n})_{n \in \mathbb{N}}$  is also a martingale with respect to  $(\mathcal{F}_n)_{n \in \mathbb{N}}$ .

### 2.3 Applications: Ruin Probability

Consider the standard random walk (or gambling process)  $(S_n)_{n \in \mathbb{N}}$  on  $\{0, 1, \dots, B\}$  with independent  $\{-1, 1\}$ -valued increments, and

$$\mathbb{P}(S_{n+1} - S_n = +1) = p \quad \text{and} \quad \mathbb{P}(S_{n+1} - S_n = -1) = q, \quad n \in \mathbb{N}$$

Let

$$T_{0,B} : \Omega \longrightarrow \mathbb{N}$$

be the first hitting time of the boundary  $\{0, B\}$ , define stopping time  $\tau$  by

$$\tau := T_{0,B} := \inf\{n \geq 0 : S_n = B \text{ or } S_n = 0\}$$

We will recover the ruin probabilities

$$\mathbb{P}(s_\tau = 0 | S_0 = k), \quad k = 0, 1, \dots, B$$

First in the unbiased case  $p = q = 1/2$

### 2.3.1 Unbiased Ruin Probability with Martingale

Step 1. The process  $(S_n)_{n \in \mathbb{N}}$  is a martingale.

We note that the process  $(S_n)_{n \in \mathbb{N}}$  has independent increments, and in the unbiased case  $p = q = 1/2$  those increments are centered:

$$\mathbb{E}[S_{n+1} - S_n] = 1 \times p + (-1) \times q = 1 \times \frac{1}{2} + (-1) \times \frac{1}{2} = 0,$$

hence  $(S_n)_{n \in \mathbb{N}}$  is a *martingale*.

Step 2. The stopped process  $(S_{\tau \wedge n})_{n \in \mathbb{N}}$  is also a martingale, as a consequence of Theorem 1.6 (Doob's stopping time theorem)

Step 3. Since the stopped process  $(S_{\tau \wedge n})_{n \in \mathbb{N}}$  is a martingale, we find that its expectation  $\mathbb{E}[S_{\tau \wedge n} | S_0 = k]$  is constant in  $n \in \mathbb{N}$  by proposition 1.2, which gives

$$k = \mathbb{E}[S_0 | S_0 = k] = \mathbb{E}[S_{\tau \wedge n} | S_0 = k], \quad k = 0, 1, \dots, B$$

Letting  $n$  go to infinity we get

$$\begin{aligned} \mathbb{E}[S_\tau | S_0 = k] &= \mathbb{E} \left[ \lim_{n \rightarrow \infty} S_{\tau \wedge n} | S_0 = k \right] \\ &= \lim_{n \rightarrow \infty} \mathbb{E}[S_{\tau \wedge n} | S_0 = k] = k \end{aligned}$$

where the exchange between limit and expectation is justified by the boundedness  $|S_{\tau \wedge n}| \leq B$  a.s.,  $n \in \mathbb{N}$ .

Hence we have

$$\begin{cases} 0 \times \mathbb{P}(S_\tau = 0 | S_0 = k) + B \times \mathbb{P}(S_\tau = B | S_0 = k) = \mathbb{E}[S_\tau | S_0 = k] = k \\ \mathbb{P}(S_\tau = 0 | S_0 = k) + \mathbb{P}(S_\tau = B | S_0 = k) = 1 \end{cases}$$

which shows that

$$\mathbb{P}(S_\tau = B | S_0 = k) = \frac{k}{B} \quad \text{and} \quad \mathbb{P}(S_\tau = 0 | S_0 = k) = 1 - \frac{k}{B}$$

### 2.3.2 Biased Ruin Probability with Martingale

Next, for the unbiased case where  $p \neq q$ . Here we note that the process

$$M_n := \left(\frac{q}{p}\right)^{S_n}, \quad n \in \mathbb{N}$$

is a martingale with respect to  $(\mathcal{F}_n)_{n \in \mathbb{N}}$ .

Step 1. The process  $(M_n)_{n \in \mathbb{N}}$  is a martingale.

$$\begin{aligned} \mathbb{E}[M_{n+1} | \mathcal{F}_n] &= \mathbb{E} \left[ \left(\frac{q}{p}\right)^{S_{n+1}} \middle| \mathcal{F}_n \right] = \mathbb{E} \left[ \left(\frac{q}{p}\right)^{S_{n+1}-S_n} \left(\frac{q}{p}\right)^{S_n} \middle| \mathcal{F}_n \right] \\ &= \left(\frac{q}{p}\right)^{S_n} \mathbb{E} \left[ \left(\frac{q}{p}\right)^{S_{n+1}-S_n} \middle| \mathcal{F}_n \right] \\ &= \left(\frac{q}{p}\right)^{S_n} \mathbb{E} \left[ \left(\frac{q}{p}\right)^{S_{n+1}-S_n} \right] \\ &= \left(\frac{q}{p}\right)^{S_n} \left( \frac{q}{p} \mathbb{P}(S_{n+1} - S_n = 1) + \left(\frac{q}{p}\right)^{-1} \mathbb{P}(S_{n+1} - S_n = -1) \right) \\ &= \left(\frac{q}{p}\right)^{S_n} \left( p \frac{q}{p} + q \left(\frac{q}{p}\right)^{-1} \right) \\ &= \left(\frac{q}{p}\right)^{S_n} (q + p) = \left(\frac{q}{p}\right)^{S_n} = M_n \end{aligned}$$

$n \in \mathbb{N}$ . In particular, the expectation of  $(M_n)_{n \in \mathbb{N}}$  is constant over time by Proposition 1.2 since it is a martingale, *i.e.* we have

$$\left(\frac{q}{p}\right)^k = \mathbb{E}[M_0 | S_0 = k] = \mathbb{E}[M_n | S_0 = k], \quad k = 0, 1, \dots, B, \quad n \in \mathbb{N}$$

Step 2. The stopped process  $(M_{\tau \wedge n})_{n \in \mathbb{N}}$  is also a martingale, as a consequence of Theorem 1.6

Step 3. Since the stopped process  $(M_{\tau \wedge n})_{n \in \mathbb{N}}$  remains a martingale, its expected value  $\mathbb{E}[M_{\tau \wedge n} | S_0 = k]$  is constant in  $n \in \mathbb{N}$  by Proposition 1.2, this gives

$$\left(\frac{q}{p}\right)^k = \mathbb{E}[M_0 | S_0 = k] = \mathbb{E}[M_{\tau \wedge n} | S_0 = k]$$

Next, letting  $n$  go to infinity we find

$$\begin{aligned} \left(\frac{q}{p}\right)^k &= \mathbb{E}[M_0 | S_0 = k] = \lim_{n \rightarrow \infty} \mathbb{E}[M_{\tau \wedge n} | S_0 = k] \\ &= \mathbb{E}[\lim_{n \rightarrow \infty} M_{\tau \wedge n} | S_0 = k] \\ &= \mathbb{E}[M_\tau | S_0 = k] \end{aligned}$$

hence

$$\begin{aligned}
\left(\frac{q}{p}\right)^k &= \mathbb{E}[M_\tau | S_0 = k] \\
&= \left(\frac{q}{p}\right)^B \mathbb{P}\left(M_\tau = \left(\frac{q}{p}\right)^B \middle| S_0 = k\right) + \left(\frac{q}{p}\right)^0 \mathbb{P}\left(M_\tau = \left(\frac{q}{p}\right)^0 \middle| S_0 = k\right) \\
&= \left(\frac{q}{p}\right)^B \mathbb{P}\left(M_\tau = \left(\frac{q}{p}\right)^B \middle| S_0 = k\right) + \mathbb{P}(M_\tau = 1 | S_0 = k)
\end{aligned}$$

Solving the system of equations

$$\begin{cases} \left(\frac{q}{p}\right)^k = \left(\frac{q}{p}\right)^B \mathbb{P}\left(M_\tau = \left(\frac{q}{p}\right)^B \middle| S_0 = k\right) + \mathbb{P}(M_\tau = 1 | S_0 = k) \\ \mathbb{P}\left(M_\tau = \left(\frac{q}{p}\right)^B \middle| S_0 = k\right) + \mathbb{P}(M_\tau = 1 | S_0 = k) = 1 \end{cases}$$

gives

$$\begin{aligned}
\mathbb{P}(S_\tau = B | S_0 = k) &= \mathbb{P}\left(M_\tau = \left(\frac{q}{p}\right)^B \middle| S_0 = k\right) \\
&= \frac{(q/p)^k - 1}{(q/p)^B - 1}
\end{aligned}$$

and

$$\begin{aligned}
\mathbb{P}(S_\tau = 0 | S_0 = k) &= \mathbb{P}(M_\tau = 1 | S_0 = k) \\
&= 1 - \frac{(q/p)^k - 1}{(q/p)^B - 1} \\
&= \frac{(q/p)^B - (q/p)^k}{(q/p)^B - 1}
\end{aligned}$$

for  $k = 0, 1, \dots, B$ .

## 2.4 Applications: Mean Game Duration

In this section we show how we can recover the mean game duration  $\mathbb{E}[\tau|S_0 = k]$ .

### 2.4.1 Unbiased Mean Game Duration with Martingale

In the case of a fair game  $p = q = 1/2$ ,

Step 1. The process  $(S_n^2 - n)_{n \in \mathbb{N}}$  is a martingale.

$$\begin{aligned}
\mathbb{E}[S_{n+1}^2 - (n+1)|\mathcal{F}_n] &= \mathbb{E}[(S_n + S_{n+1} - S_n)^2 - (n+1)|\mathcal{F}_n] \\
&= \mathbb{E}[S_n^2 + (S_{n+1} - S_n)^2 + 2S_n(S_{n+1} - S_n) - (n+1)|\mathcal{F}_n] \\
&= \mathbb{E}[S_n^2 - n - 1|\mathcal{F}_n] + \mathbb{E}[(S_{n+1} - S_n)^2|\mathcal{F}_n] + 2\mathbb{E}[S_n(S_{n+1} - S_n)|\mathcal{F}_n] \\
&= S_n^2 - n - 1 + \mathbb{E}[(S_{n+1} - S_n)^2] + 2S_n\mathbb{E}[S_{n+1} - S_n] \\
&= S_n^2 - n - 1 + \mathbb{E}[(S_{n+1} - S_n)^2] \\
&= S_n^2 - n
\end{aligned}$$

Step 2. The stopped process  $(S_{\tau \wedge n}^2 - \tau \wedge n)_{n \in \mathbb{N}}$  is also a martingale, as a consequence of Theorem 1.6

Step 3. Since the stopped process  $(S_{\tau \wedge n}^2 - \tau \wedge n)_{n \in \mathbb{N}}$  is also a martingale, its expectation  $\mathbb{E}[S_{\tau \wedge n}^2 - \tau \wedge n|S_0 = k]$  is constant in  $n \in \mathbb{N}$  by Proposition 1.2, hence we have

$$k^2 = \mathbb{E}[S_0^2 - 0|S_0 = k] = \mathbb{E}[S_{\tau \wedge n}^2 - \tau \wedge n|S_0 = k]$$

and after taking the limit as  $n$  tends to infinity,

$$\begin{aligned}
k^2 &= \lim_{n \rightarrow \infty} \mathbb{E}[S_{\tau \wedge n}^2 - \tau \wedge n|S_0 = k] \\
&= \lim_{n \rightarrow \infty} \mathbb{E}[S_{\tau \wedge n}^2|S_0 = k] - \lim_{n \rightarrow \infty} \mathbb{E}[\tau \wedge n|S_0 = k] \\
&= \mathbb{E}\left[\lim_{n \rightarrow \infty} S_{\tau \wedge n}^2|S_0 = k\right] - \mathbb{E}\left[\lim_{n \rightarrow \infty} \tau \wedge n|S_0 = k\right] \\
&= \mathbb{E}\left[\lim_{n \rightarrow \infty} S_{\tau \wedge n}^2 - \lim_{n \rightarrow \infty} \tau \wedge n|S_0 = k\right] \\
&= \mathbb{E}[S_\tau^2 - \tau|S_0 = k]
\end{aligned}$$

since  $S_{\tau \wedge n}^2 \in [0, B^2]$  for all  $n \in \mathbb{N}$  and  $n \mapsto \tau \wedge n$  is nondecreasing, and this gives

$$\begin{aligned}
k^2 &= \mathbb{E}[S_\tau^2 - \tau|S_0 = k] \\
&= \mathbb{E}[S_\tau^2|S_0 = k] - \mathbb{E}[\tau|S_0 = k] \\
&= B^2\mathbb{P}(S_\tau = B|S_0 = k) + 0^2\mathbb{P}(S_\tau = 0|S_0 = k) - \mathbb{E}[\tau|S_0 = k]
\end{aligned}$$

*i.e.*

$$\begin{aligned}
\mathbb{E}[\tau|S_0 = k] &= B^2(S_\tau = B|S_0 = k) - k^2 \\
&= B^2 \frac{k}{B} - k^2 \\
&= k(B - k), \quad k = 0, 1, \dots, B
\end{aligned}$$

### 2.4.2 Biased Mean Game Duration with Martingale

In the case of non-symmetric case where  $p \neq q$

Step 1. The process  $S_n - (p - q)n$  is a martingale.

$$\mathbb{E}[S_n - S_{n-1} - (p - q)] = \mathbb{E}[S_n - S_{n-1}] - (p - q) = 0$$

Step 2. The stopped process  $(S_{\tau \wedge n} - (p - q)(\tau \wedge n))_{n \in \mathbb{N}}$  is also a martingale, as a consequence of Theorem 1.6

Step 3. The expectation  $\mathbb{E}[S_{\tau \wedge n} - (p - q)(\tau \wedge n) | S_0 = k]$  is constant in  $n \in \mathbb{N}$

Step 4. Since the stopped process  $(S_{\tau \wedge n} - (p - q)(\tau \wedge n))_{n \in \mathbb{N}}$  is a martingale, we have

$$k = \mathbb{E}[S_0 - 0 | S_0 = k] = \mathbb{E}[S_{\tau \wedge n} - (p - q)(\tau \wedge n) | S_0 = k]$$

and after taking the limit as  $n$  tends to infinity,

$$\begin{aligned} k &= \lim_{n \rightarrow \infty} \mathbb{E}[S_{\tau \wedge n} - (p - q)(\tau \wedge n) | S_0 = k] \\ &= \mathbb{E}[\lim_{n \rightarrow \infty} S_{\tau \wedge n} - (p - q) \lim_{n \rightarrow \infty} \tau \wedge n | S_0 = k] \\ &= \mathbb{E}[S_\tau - (p - q)\tau | S_0 = k] \end{aligned}$$

which gives

$$\begin{aligned} k &= \mathbb{E}[S_\tau - (p - q)\tau | S_0 = k] \\ &= \mathbb{E}[S_\tau | S_0 = k] - (p - q)\mathbb{E}[\tau | S_0 = k] \\ &= B \times \mathbb{P}(S_\tau = B | S_0 = k) + 0 \times \mathbb{P}(S_\tau = 0 | S_0 = k) - (p - q)\mathbb{E}[\tau | S_0 = k] \end{aligned}$$

*i.e.*

$$\begin{aligned} (p - q)\mathbb{E}[\tau | S_0 = k] &= B \times \mathbb{P}(S_\tau = B | S_0 = k) - k \\ &= B \frac{(q/p)^k - 1}{(q/p)^B - 1} - k, \end{aligned}$$

hence

$$\mathbb{E}[\tau | S_0 = k] = \frac{1}{p - q} \left( B \frac{(q/p)^k - 1}{(q/p)^B - 1} - k \right)$$

for  $k = 0, 1, \dots, B$

## 2.5 Applications: Summary

Here's the summarized table of the family of martingales.

Table 1.1: List of martingales.

	Unbiased	Biased
Ruin probability	$S_n$	$\left(\frac{q}{p}\right)^{S_n}$
Mean game duration	$S_n^2 - n$	$S_n - n(p - q)$



## 3 (2) Assets, Portfolios, and Arbitrage

### 3.1 (0) Back to the Introduction

#### Definition 0.1: Put option

A (European) *put* option is a contract that gives its holder the right to *sell* a quantity of assets at a predefined price  $K$  called the strike price and at a predefined date  $T$  called maturity

In general, the payoff of a put option takes the form

$$\phi(S_T) = (K - S_T)^+ := \begin{cases} K - S_T & S_T \leq K \\ 0 & S_T \geq K \end{cases}$$

#### Definition 0.2: Call option

A (European) *call* option is a contract that gives its holder the right to *buy* a quantity of assets at a predefined price  $K$  called the strike price and at a predefined date  $T$  called maturity

In general, the payoff of a put option takes the form

$$\phi(S_T) = (S_T - K)^+ := \begin{cases} 0 & S_T \leq K \\ S_T - K & S_T \geq K \end{cases}$$

#### Definition 0.3: Pricing and hedging in a binary model

The *arbitrage price* of the option is interpreted as the initial cost  $\alpha S_0 + \$\beta$  of the portfolio hedging the claim  $\mathcal{C}$

## 3.2 (2.4) Risk-Neutral Probability Measures

Bunch of theorem below...

### Theorem 2.5

A market is *without* arbitrage opportunity  $\iff$  It admits at least one (equivalent<sup>a</sup>) risk-neutral probability measure  $\mathbb{P}^*$

<sup>a</sup>Referring to Definition 2.4; A probability measure  $\mathbb{P}^*$  on  $(\Omega, \mathcal{F})$  is said to be *equivalent* to another probability measure  $\mathbb{P}$  when  $\mathbb{P}^*(A) = 0 \iff \mathbb{P}(A) = 0$  for all  $A \in \mathcal{F}$

## 3.3 (2.6) Market Completeness

### Definition 2.10: Pricing and hedging in a binary model

A market model is said to be *complete* if every contingent claim  $\mathcal{C}$  is attainable.

### Definition 2.11: Pricing and hedging in a binary model

A market model without arbitrage opportunities is complete  $\iff$  It admits only one (equivalent) risk-neutral probability measure  $\mathbb{P}^*$ .

## 4 (3) Discrete-Time Model

### 4.1 (3.5) Multingales and Conditional Expectation

#### Definition 3.4

A stochastic process  $(M_t)_{t=0,1,\dots,N}$  is called a discrete-time martingale with respect to the filtration  $(\mathcal{F}_t)_{t=0,1,\dots,N}$  if  $(M_t)_{t=0,1,\dots,N}$  is  $(\mathcal{F}_t)_{t=0,1,\dots,N}$ -adapted and satisfies the property

$$\mathbb{E}[M_{t+1}|\mathcal{F}_t] = M_t, \quad t = 0, 1, \dots, N-1.$$

#### Proposition 3.5

Let  $(Z_n)_{n \in \mathbb{N}}$  be martingale, we have

$$\mathbb{E}[Z_n] = \mathbb{E}[Z_0], \quad n \in \mathbb{N}$$

### 4.2 (3.6) Market Completeness and Risk-Neutral Measures

#### Definition 3.8

A probability measure  $\mathbb{P}^*$  on  $\Omega$  is called a risk-neutral measure if

$$\mathbb{E}^* \left[ S_{t+1}^{(i)} \middle| \mathcal{F}_t \right] = (1+r)S_t^{(i)}, \quad i = 1, 2, \dots, d$$

#### Definition 3.9

A probability measure  $\mathbb{P}^*$  on  $\Omega$  is called a risk-neutral measure if

$$\tilde{S}_t^{(i)} := \frac{S_t^{(i)}}{(1+r)^t}, \quad t = 0, 1, \dots, n$$

is a martingale under  $\mathbb{P}^*$ , i.e.

$$\mathbb{E}^* \left[ \tilde{S}_{t+1}^{(i)} \middle| \mathcal{F}_t \right] = \tilde{S}_t^{(i)}, \quad t = 0, 1, \dots, N-1$$

$i = 0, 1, \dots, d.$

### 4.3 (3.7) The Cox-Ross-Rubinstein (CRR) Market Model

Consider the portfolio consists of

- Risk-free asset priced as

$$S_t^{(0)} = S_0^{(0)}(1+r)^t, \quad t = 0, 1, \dots, N$$

- Risky asset priced as

$$S_t^{(1)} = S_0^{(1)} \prod_{k=1}^t (1 + R_k), \quad t = 0, 1, \dots, N$$

where  $R_t \in \{a, b\}$ ,  $t = 1, 2, \dots, N$

#### Theorem 3.15

The CRR model is without arbitrage opportunities if and only if  $a < r < b$ . In this case the market is complete and the (equivalent) risk-neutral probability measure  $\mathbb{P}^*$  is given by

$$\mathbb{P}^*(R_{t+1} = b | \mathcal{F}_t) = \frac{r - a}{b - a} \quad \text{and} \quad \mathbb{P}^*(R_{t+1} = a | \mathcal{F}_t) = \frac{b - r}{b - a}$$

$t = 0, 1, \dots, N - 1$ . In particular,  $(R_1, R_2, \dots, R_N)$  forms a sequence of *i.i.d* random variables under  $\mathbb{P}^*$ , with

$$p^* := \mathbb{P}^*(R_t = b) = \frac{r - a}{b - a} \quad \text{and} \quad q^* := \mathbb{P}^*(R_t = a) = \frac{b - r}{b - a}$$

## 5 (4) Pricing and Hedging in Discrete Time

### 5.1 (4.1) Pricing of Contingent Claims

#### Lemma 4.2

The following statements are equivalent:

- (i) The portfolio strategy  $(\bar{\xi}_t)_{t=1,2,\dots,N}$  is self-financing
- (ii)  $\bar{\xi}_t \cdot \bar{X}_t = \bar{\xi}_{t+1} \cdot \bar{X}_t$  for all  $t = 1, 2, \dots, N - 1$ .
- (iii) The discounted portfolio value  $\tilde{V}_t$  can be written as the stochastic summation

$$\tilde{V}_t = \tilde{V}_0 + \underbrace{\sum_{k=1}^t \bar{\xi}_k \cdot (\bar{X}_k - \bar{X}_{k-1})}_{\text{sum of profits and losses}}, \quad t = 0, 1, \dots, N,$$

of discounted profits and losses.

#### Theorem 4.5

The arbitrage price  $\pi_t(C)$  of an attainable contingent claim  $C$  is given by

$$\pi_t(C) = \frac{1}{(1+r)^{N-t}} \mathbb{E}^*[C \mid \mathcal{F}_t], \quad t = 0, 1, \dots, N$$

where  $\mathbb{P}^*$  denotes any risk-neutral probability measure.

## 5.2 (4.2) Pricing of Vanilla Options in the CRR Model

### Proposition 4.8

The price  $\pi_t(C)$  of the contingent claim  $C = f(S_N^{(1)})$  satisfies

$$\pi_t(C) = v(t, S_t^{(1)}), \quad t = 0, 1, \dots, N$$

where the function  $v(t, x)$  is given by

$$\begin{aligned} v(t, x) &= \frac{1}{(1+r)^{N-t}} \mathbb{E}^* \left[ f \left( x \prod_{j=t+1}^N (1+R_j) \right) \right] \\ &= \frac{1}{(1+r)^{N-t}} \sum_{k=0}^{N-t} \binom{N-t}{k} (p^*)^k (1-p^*)^{N-t-k} f(x(1+b)^k (1+a)^{N-t-k}) \end{aligned}$$

### Useful note; Binomial Identity and Coefficient

- Binomial Identity

$$(x+y)^n = \sum_{k=0}^n \binom{n}{k} x^{n-k} y^k$$

- Binomial Coefficient

$$\binom{n}{k} = \frac{n!}{k!(n-k)!}$$

## 6 (5.4) Ito Stochastic Integral: Ito Isometry

Proposition 5.9; Ito Isometry

$$\text{Var} \left[ \int_0^T f(t) dB_t \right] = \mathbb{E} \left[ \left( \int_0^T f(t) dB_t \right)^2 \right] = \int_0^T |f(t)|^2 dt$$

Application of Ito Isometry

- i.e.1)

$$\mathbb{E} \left[ \left( \int_0^T B_t dB_t \right)^2 \right] = \mathbb{E} \left[ \int_0^T |B_t|^2 dt \right] = \int_0^T \mathbb{E} [|B_t|^2] dt = \int_0^T t dt = \frac{T^2}{2}$$

- i.e.2) For all square-integrable adapted processes  $(u_t)_{t \in \mathbb{R}_+}$ ,  $(v_t)_{t \in \mathbb{R}_+}$

$$\mathbb{E} \left[ \int_0^T u_t dB_t \int_0^T v_t dB_t \right] = \mathbb{E} \left[ \int_0^T u_t v_t dt \right]$$