

Revision notes - MA4269

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Contents

0	Review	2
1	Brownian Motion	4
2	Black-Scholes Model	6
3	Martingale & Girsanov	7
4	Change of Numeraire	9
5	American Options	10
6	Barrier Option	11
7	Asian Options	13
8	Lookback Options & Two Assets Options	15

0 Review

0.1 Introduction

Definition 0.1 (Zero-coupon Bond).

A contract to deliver \$1 on a future date T is known as a **zero-coupon bond**.

The price of the bond at time $t < T$ is denoted by $P(t, T)$, where T is the maturity of the bond.

Definition 0.2 (Money Market Account).

The **money market account** is an asset created by the following procedure, called continuously compounded:

- The initial amount equal to \$1 is invested at time $t = 0$ in the bond with the shortest available maturity (i.e. the next infinitesimal instant).
- The position is rolled over to the bond with the next shortest maturity once the first bond expires.

The price of the money market at time t is denoted by M_t .

Theorem 0.1.

We have

$$dM_t = r(t)M_t dt$$

to describe the value of money market account, where $r(t)$ is a parameter known as the interest rate.

Solving, we have

$$M_t = \exp\left(\int_0^t r(u) du\right)$$

Remark: The relationship between M_t and $P(t, T)$ is non-trivial if $r(t)$ is stochastic. However,

Theorem 0.2.

Suppose interest rate r is constant. Then,

$$P(t, T) = \frac{M_t}{M_T} = e^{r(T-t)}$$

It is obvious that $P(t, T) < 1$ as long as $r > 0$.

Definition 0.3 (Value of Asset).

The **value** of an asset is the amount of dollars that an investor will pay to own that asset.

The term **stock price** refers to the value (per unit) of the stock under consideration.

Definition 0.4 (Position).

A **position** in an asset is the quantity of an asset owned or owed by an investor.

- **long position:** the investor *owns* the asset.
- **short position:** the investor *sells* the asset that he does not own.

Definition 0.5 (Portfolio).

A **portfolio** is a combination of various positions in financial assets.

At any time t , the **value of the portfolio** Π_t is just the sum of the values of all the positions held in the portfolio at that particular time t :

$$\Pi_t = a_t^{(1)}A_t^{(1)} + \dots + a_t^{(n)}A_t^{(n)}$$

where $a_t^{(i)}$ is the position where $A_t^{(i)}$ is the price of the asset at time t .

Since the price of assets is not determined by us, therefore, Π_t can be written as the vector

$$(a_t^{(1)}, \dots, a_t^{(n)})$$

Definition 0.6 (Self-financing).

A portfolio $\Pi_t = (a_t^{(1)}, \dots, a_t^{(n)})$ is **self-financing** over the time interval $[0, T]$ if there is *no* exogenous infusion or withdrawal of money, *except* possibly at the initiation time 0 or maturity date T :

$$d\Pi_t = a_t^{(1)}dA_t^{(1)} + \dots + a_t^{(n)}dA_t^{(n)}$$

Roughly speaking, the differential equation says that the change in portfolio is completely due to the change in the underlying asset prices and nothing else.

Definition 0.7 (Direction of Cash Flow).

Suppose I am an investor

- If cash flow is positive, it means that someone pays me.
- If cash flow is negative, it means that I pay someone.

Cash flow depends on

1. the value of the underlying portfolio, and
2. whether or not the portfolio is entered into or liquidated.

0.2 Financial Market

Definition 0.8 (Arbitrage).

An **arbitrage opportunity** is the existence of a self-financing portfolio $\Pi_t, 0 \leq t \leq T$, having the following properties:

1. $\Pi_0 = 0$
2. $\Pi_T \geq 0$ for **all possible** outcomes
3. There is a positive probability that $\Pi_T > 0$.

Definition 0.9 (Equivalent Definition of Arbitrage).

An arbitrage opportunity is the existence of a self-financing portfolio $\Pi_t, 0 \leq t \leq T$, having the following properties:

1. $\Pi_T - \Pi_0 e^{rT} \geq 0$ for **all possible** outcomes
2. There is a positive probability that $\Pi_T - \Pi_0 e^{rT} > 0$

Theorem 0.3 (Consequence of No Arbitrage).

Suppose there are two self-financing portfolios Π_t^A and Π_t^B over the time interval $[t, T]$ such that $\Pi_T^A \geq \Pi_T^B$. Then in the absence of arbitrage, we must have

$$\Pi_t^A \geq \Pi_t^B$$

Theorem 0.4 (Law of One Price).

Law of one price is a consequence of the above theorem.

Suppose there are two self-financing portfolios Π_t^A and Π_t^B over the time interval $[t, T]$ such that $\Pi_T^A = \Pi_T^B$. Then, in the absence of arbitrage, we must have

$$\Pi_t^A = \Pi_t^B$$

Theorem 0.5.

All *risk-free* portfolios must earn the same return, i.e., *riskless interest rate*. Suppose Π_t is the value of a riskfree portfolio, and $d\Pi_t$ is the price increment during a small period of time interval $[t, t + dt]$. Then

$$d\Pi_t = r\Pi_t dt$$

where r is the riskless interest rate.

0.3 Forward Contracts & Options

Definition 0.10 (Forward Contract).

A **forward contract** is a contract that delivers one unit of the underlying asset on a known future date T for a certain price K agreed today.

Here,

- K is the **delivery price**
- T is called the **delivery date**
- the buyer of the contract is in the long position
- the seller of the contract is in the short position
- the delivery price K is the amount the long side pays the short side in exchange of one unit of the **underlying asset whose value** is S_T on the delivery date T .

Definition 0.11 (Forward Price).

The **forward price** at time t is the delivery price of a forward contract which costs nothing to enter into at time t . We denote the forward price at time t by

$$F(S_t, t, T)$$

Remark: The forward price $F(S, t, T)$ is *not* the value of corresponding forward contract.

Definition 0.12 (Payoff, Profit).

The **payoff** to a position is the value of the position at the maturity date T .

The **profit** to a position is the payoff to the position at maturity date T , subtracted by the time- T value of the initial investment in the position:

$$\Pi_T - \Pi_t e^{r(T-t)}$$

Theorem 0.6 (Payoff of Forward Contract).

It is obvious from the definition that

- the payoff to a long forward contract is $S_T - K$;
- the payoff to a short forward contract is $K - S_T$

Suppose it costs nothing to enter into a forward contract, then by using the forward price definition,

- the payoff and the profit to a long forward contract are the same:

$$S_T - F(S, t, T)$$

- the payoff and profit to a short forward contract are the same:

$$F(S, t, T) - S_T$$

Theorem 0.7 (Forward Price).

Suppose the underlying stock S does not pay dividends. Then the forward price $F(S, t, T)$ of stock at time t is given by

$$F(S, t, T) = S e^{r(T-t)}$$

where S is the price of the stock at time t .

Definition 0.13 (Call Option).

A **call option** is an agreement where the buyer has the *right*, but not the obligation to buy the underlying asset, for a certain price K agreed at the initiation of the contract. Here, K is called the strike price, whereby T is used to denote maturity, which is the date by which option must be exercised or it becomes worthless.

For now, we consider only European call option, where exercise of the contract occurs only at maturity T .

The payoff to a long European call option with strike price K and maturity T is

$$(S_T - K)^+ = \max\{S_T - K, 0\}$$

where the payoff to a short European call option with same strike price and maturity is $-(S_T - K)^+$.

Definition 0.14 (Put Option).

A **put option** is an agreement where the buyer has the right to sell an asset, but not the obligation to sell, for a certain price K agreed at the initiation of the contract.

The payoff to a long European put option with strike price K and expiration T is

$$(K - S_T)^+ = \max\{K - S_T, 0\}$$

whereas the payoff to a short European put option with same strike price and expiration T is $-(K - S_T)^+$.

Definition 0.15 (Moneyness).

Options are often described by their degree of moneyness.

At any time t , an option is said to be

- **in-the-money** if payoff at time $t > 0$.
- **at-the-money** if payoff = 0, i.e., $S_t = K$.
- **out-of-the-money** if payoff < 0.

Theorem 0.8 (Put Call Parity).

We have the following relationship between call c and put p price, over the underlying asset at time t . Here K is the strike price of the options and F is the forward price at time t .

$$c - p + (K - F)e^{-r(T-t)} = 0$$

0.4 Binomial Model

Definition 0.16 (One-period Binomial Model).

Suppose the non-dividend paying stock price per share today is S_0 . We assume that, at the end of the one period, the stock price is either S_0u or S_0d where d and u are positive real numbers such that $d < u$.

We call u the **up factor** and d the **down factor**.

Consider a derivative on the stock with time T to maturity. Let V_0 be the price of derivative at time 0.

We can price V_0 by constructing $\Pi_0 = V_0 - \phi S_0$, and make it riskless, i.e. $\Pi_T = V_u - \phi S_0u = V_d - \phi S_0d$, by picking a suitable ϕ . Since Π_0 is riskless, its payoff should be the same as any other riskless payoff, e.g. money market account, i.e., $\Pi_T = (V_0 - \phi S_0)e^{rT}$.

Solving, we have

$$V_0 = e^{-rT}(pV_u + (1 - p)V_d)$$

where $p = \frac{e^{rT} - d}{u - d}$.

We can interpret p and $1 - p$ as probabilities distribution on S_T , so that we can write

$$V_0 = e^{-rT}E^{\mathbb{Q}}[V_T]$$

The expectation of S_T under \mathbb{Q} is

$$E^{\mathbb{Q}}[S_T] = S_0e^{rT}$$

which matches our riskless argument.

Theorem 0.9 (Restriction on u and d).

In the one-period binomial model where the one period is $[0, T]$ and the corresponding up-factor and down-factor of a non-dividend paying stock are u and d respectively with $d < u$, we have

$$d < e^{rT} < u$$

Definition 0.17 (Multi-period Binomial Model).

At any time $j\Delta t$, there are $j + 1$ possible stock prices:

$$S_0d^j, S_0d^{j-1}u, \dots, S_0u^j$$

Without loss of generality, we can assume that $ud = 1$.

If V_j^k is the price of the derivative at time $j\Delta t$ when the underlying stock price is $S_0d^{j-k}u^k$, i.e. there are k period out of j that the price goes up.

We then have

$$V_0 = e^{-rn\Delta t} \sum_{i=0}^n \binom{n}{i} p^i (1-p)^{n-i} V_n^i$$

where V_0 is the price of the European style derivative given by the n period binomial model.

1 Brownian Motion

1.1 Brownian Motion

Definition 1.1 (Standard Brownian Motion).

A **standard Brownian Motion** is a stochastic process $W_t, t \geq 0$ with the following defining characteristics:

(W1) $W_0 = 0$.

(W2) With probability 1 (almost surely), the function $t \rightarrow W_t$ is continuous in t .

(W3) For every $0 \leq t_1 < t_2$, $W_{t_2} - W_{t_1}$ is normally distributed with mean 0 and variance $t_2 - t_1$.

(W4) $W_{t_3} - W_{t_2}$ is independent of $W_{t_1} - W_{t_0}$ for any $0 \leq t_0 \leq t_1 \leq t_2 \leq t_3$, i.e. non-overlapping increments are independently distributed.

Property (W3) implies that for any Δt , $W_{t+\Delta t} - W_t \sim N(0, \Delta t)$. This implies that $|W_t| \leq 1.96\sqrt{t}$ with 95% probability.

Property (W4) implies that $\text{Cov}(W_t, W_s) = E[W_t W_s] = \min\{t, s\}$.¹

Theorem 1.1 (Binomial Approximation to Brownian Motion).

Let ε_1, \dots be a sequence of independent, identically distributed random variables with mean 0 and variance 1. For

¹We write $W_t = (W_t - W_s + W_s)$ if $t > s$.

each $n \geq 1$, define a continuous time stochastic process $W_t^{(n)}$ by

$$W_t^{(n)} = \frac{1}{\sqrt{n}} \sum_{1 \leq i \leq [nt]} \varepsilon_i$$

$W_t^{(n)}$ approaches a standard Brownian motion $N(0, t)$ as $n \rightarrow \infty$.

Theorem 1.2 (Infinitesimal Brownian Increment).

We can approximate dW_t in a very small time interval Δt :

$$\Delta W_t := W_{t+\Delta t} - W_t$$

and then we have

$$\Delta W_t = \phi \sqrt{\Delta t} \text{ i.e., } \Delta W_t \sim N(0, \Delta t)$$

where ϕ has a standard normal distribution.

Definition 1.2 (Generalised Wiener Process).

A **generalised Wiener Process** for a variable X can be defined in terms of dW_t as

$$dX_t = a dt + b dW_t$$

where a and b are constants. The parameter a is called the **drift rate** and b^2 is called the **variance rate** of the process.

In a small time interval Δt , the change ΔX_t is given by

$$\Delta X_t = a\Delta t + b\Delta W_t$$

Therefore, ΔX_t has a normal distribution with mean $a\Delta t$ and variance $b^2\Delta t$.

Here, we can safely write $X_t = X_0 + at + bW_t$.

1.2 Quadratic Variation

Definition 1.3 (Quadratic Variation). *Any sequence of values $0 = t_0 < t_1 < \dots < t_n = T$ is called a partition $\Pi = \Pi(t_0, \dots, t_n)$ of a fixed interval $[0, T]$. The discrete quadratic variation of a standard Brownian motion W relative to the partition Π is defined as*

$$Q(W, \Pi) = \sum_{i=1}^n (W_{t_i} - W_{t_{i-1}})^2$$

For any partition Π , define

$$\|\Pi\| = \max_{1 \leq i \leq n} |t_i - t_{i-1}|$$

Theorem 1.3 (n th Moment of Standard Normal Z).

The n th moment of a random variable X is defined to be $E[X^n]$. If $\phi \sim N(0, 1)$, then

$$E[\phi^n] = \begin{cases} 0 & \text{if } n \text{ is odd} \\ \frac{(2k)!}{2^k k!} & \text{if } n = 2k \end{cases}$$

In particular, we have $E(\phi^2) = 1$ and $E(\phi^4) = 3$.

Theorem 1.4.

Consider an arbitrary sequence of partitions Π_n , where $n = 1, 2, \dots$. Suppose $\lim_{n \rightarrow \infty} \|\Pi_n\| = 0$, then

$$\lim_{n \rightarrow \infty} E(Q(W, \Pi_n) - T)^2 = 0$$

That is the standard Brownian motion has quadratic variation which is equal to T , in the mean square limit.

Definition 1.4.

Define the integral $\int_0^T (dW)^2$ by

$$\lim_{n \rightarrow \infty} E\left[\sum_{i=1}^n (W_{t_i} - W_{t_{i-1}})^2 - \int_0^T (dW_t)^2\right]^2 = 0$$

However, from quadratic variation theorem, we have

$$\lim_{n \rightarrow \infty} E\left[\sum_{i=1}^n (W_{t_i} - W_{t_{i-1}})^2 - \int_0^T dt\right]^2 = 0$$

Therefore,

$$\int_0^T (dW_t)^2 = \int_0^T dt$$

In fact, we can write

$$(dW_t)^2 = dt$$

which gives

$$(\Delta W)^2 \approx \Delta t$$

in discrete time approximation.

1.3 Itô's lemma

Definition 1.5 (Itô's Process).

The Itô's Process dX_t is defined as

$$dX_t = a(X_t, t) dt + b(X_t, t) dW_t$$

Theorem 1.5 (Itô's Lemma).

Let $V(X_t, t)$ be a smooth function of t and of the Itô's process X_t :

$$dX_t = a dt + b dW_t$$

for some $a = a(X_t, t)$ and $b = b(X_t, t)$. Then we have

$$dV(X_t, t) = \left(a \frac{\partial V}{\partial X} + \frac{\partial V}{\partial t} + \frac{1}{2} b^2 \frac{\partial^2 V}{\partial X^2}\right) dt + b \frac{\partial V}{\partial X} dW_t$$

Another version of the Itô's Lemma where we do not have explicit form of dX is

$$dV_t = dV(X_t, t) = \frac{\partial V}{\partial t} dt + \frac{\partial V}{\partial X} dX_t + \frac{1}{2} \frac{\partial^2 V}{\partial X^2} (dX_t)^2$$

Definition 1.6 (Ito Integral).

In order to have differential form of the SDE $dX_t = a(X_t, t) dt + b(X_t, t) dW_t$, we require $a(X_t, t)$ and $b(X_t, t)$ to be non-anticipative, which means that its value at t can only be available at time t .

With the above assumption, we can define

$$\int_0^T a(X_t, t) dt = \lim_{n \rightarrow \infty} \sum_{i=1}^n a(X_{t_{i-1}}, t_{i-1})(t_i - t_{i-1})$$

and Ito integral

$$\int_0^T b(X_t, t) dW_t$$

is the mean-square limit of the sum $\sum_{i=1}^n b(X_{t_{i-1}}, t_{i-1})(W_{t_i} - W_{t_{i-1}})$.

2 Black-Scholes Model

In Black-Scholes Model, we assume the following 2 conditions:

1. The money market(riskless asset) M_t is given by

$$dM_t = rM_t dt$$

2. The stock price follows the Geometric Brownian motion:

$$dS_t = \mu S_t dt + \sigma S_t dW_t$$

where μ and σ are constants.

We can derive Black Scholes PDE from either delta hedging, where we take $\Pi_t = V_t - \phi_t S_t$, $0 \leq t \leq T$, such that ϕ_t is chosen to make Π_t self-financing and riskless. Self financing condition gives

$$d\Pi_t = dV_t - \phi_t dS_t$$

Also, since ϕ_t is chosen so that Π_t is riskless, we also need

$$d\Pi_t = r\Pi_t dt$$

Here, dV_t can be calculated via Ito's lemma and dS_t is given in assumption. Solving, we will have

$$\phi_t = \frac{\partial V}{\partial S}$$

and

$$\frac{\partial V}{\partial t} + r \frac{\partial V}{\partial S} S_t + \frac{1}{2} \sigma^2 S_t^2 \frac{\partial^2 V}{\partial S^2} = rV$$

Here, we call $\frac{\partial V}{\partial S}$ **delta** of the derivative.

We can derive the Black Scholes PDE by *replication* also, by considering an asset $\Pi_t = a_t S_t + b_t M_t$, which satisfies $\Pi_t = V_t$ for all $t \leq T$. The equality throughout T gives

$$d\Pi_t = dV_t.$$

Similarly, self financing condition gives

$$d\Pi_t = a_t dS_t + b_t dM_t$$

where dS_t and dM_t are readily available.

Also, we can compute dV_t via Ito's Lemma:

$$dV_t = \left(\frac{\partial V}{\partial S} \mu S_t + \frac{\partial V}{\partial t} + \frac{1}{2} \frac{\partial^2 V}{\partial S^2} \sigma^2 S_t^2 \right) dt + \frac{\partial V}{\partial S} \sigma S_t dW_t$$

Then we compare coefficients of dW_t , arriving at

$$a_t = \frac{\partial V}{\partial S}$$

and therefore b_t can be written as

$$b_t = \frac{1}{M_t} (V_t - \frac{\partial V}{\partial S} S_t)$$

whereas comparing dt and make necessary computation, we can arrive at

$$\frac{\partial V}{\partial t} + r \frac{\partial V}{\partial S} S_t + \frac{1}{2} \sigma^2 S_t^2 \frac{\partial^2 V}{\partial S^2} - rV = 0$$

Remark:

1. The drift parameter μ of the stock never enters into the PDE.
2. To uniquely determine the solution, we must prescribe
 - Boundary conditions
 - Initial, or final conditions

Theorem 2.1 (Solution to Black Scholes PDE).

For a European call option with a call price $c(S_t, t)$, we can make the following observation:

1. Final condition: $c(S_T, T) = \max\{S_T - K, 0\}$
2. Boundary condition 1: $S_0 = 0 \Rightarrow c(0, t) = 0$ for all $0 \leq t \leq T$.
3. Boundary condition 2: With $S_t \gg K$, we have $c(S_t, t) \approx S_t$ for all $0 \leq t \leq T$.

With these observation, we have, for European call

$$c_t = S_t N(d_+) - K e^{-r\tau} N(d_-)$$

For European put:

$$p_t = K e^{-r\tau} N(-d_-) - S_t N(-d_+)$$

where

$$d_{\pm} = \frac{\ln(S_t/K) + (r \pm \sigma^2/2)\tau}{\sigma \sqrt{\tau}}$$

and

$$\tau = T - t$$

Theorem 2.2 (Black Scholes PDE with Presence of Dividends).

With presence of dividends,

$$d\Pi_t = a_t dS_t + b_t dM_t + a_t q S_t dt$$

The black scholes PDE becomes

$$\frac{\partial V}{\partial t} + (r - q) \frac{\partial V}{\partial S} S_t + \frac{1}{2} \sigma^2 S_t^2 \frac{\partial^2 V}{\partial S^2} - rV = 0$$

2.1 Preliminaries on Martingale Pricing

Definition 2.1 (Equivalent Probability Measure).

Suppose there are two probability measures \mathbb{P} and \mathbb{Q} on space (Ω, \mathcal{F}) . We say that \mathbb{P} and \mathbb{Q} are **equivalent**, denoted by $\mathbb{P} \sim \mathbb{Q}$ if

$$\mathbb{P}(A) > 0 \Leftrightarrow \mathbb{Q}(A) > 0 \text{ for all } A \in \mathcal{F}$$

Essentially, two equivalent measures agree on **all** certain and impossible events.

We hope to derive an equivalent probability measure \mathbb{Q} from an existing one \mathbb{P} . To do this, we require a **positive** random variable L with property $\mathbb{E}^{\mathbb{P}}[L] = 1$. These two conditions are two defining characteristic of a Radon-Nikodym Derivative L .

Define \mathbb{Q} by

$$\mathbb{Q}(A) = \mathbb{E}^{\mathbb{P}}[L \cdot \mathbb{1}_A]$$

where $\mathbb{1}_A$ is the indicator random variable for the event A .

Theorem 2.3 (Radon Nikodym).

Consider two probability measures \mathbb{P} and \mathbb{Q} on (Ω, \mathcal{F}) . The following are equivalent:

1. $\mathbb{P} \sim \mathbb{Q}$
2. There exists a positive random variable L such that for every event $A \in \mathcal{F}$

- $\mathbb{Q}(A) = \mathbb{E}^{\mathbb{P}}[L \cdot \mathbf{1}_A]$
- $\mathbb{P}(A) = \mathbb{E}^{\mathbb{Q}}[\frac{1}{L} \cdot \mathbf{1}_A]$.

Here $L = \frac{d\mathbb{Q}}{d\mathbb{P}}$, as derived from theorem, is called the Radon Nikodym derivative of \mathbb{Q} wrt \mathbb{P} .

Theorem 2.4.

Let X be a random variable. With above notations, we have

$$\mathbb{E}^{\mathbb{Q}}[X] = \mathbb{E}^{\mathbb{P}}[L \cdot X]$$

and

$$\mathbb{E}^{\mathbb{P}}[X] = \mathbb{E}^{\mathbb{Q}}[\frac{1}{L} \cdot X]$$

Definition 2.2 (Filtration).

Let $\{\mathcal{F}_t\}, 0 \leq t \leq T$ be a filtration, where \mathcal{F}_t is information available to us at time t .

Trivially, we have $\mathcal{F}_s \subseteq \mathcal{F}_t$ for $s \leq t$.

Suppose we now want to move from $(\Omega, \{\mathcal{F}_t\}, \mathbb{P})$ to $(\Omega, \{\mathcal{F}_t\}, \mathbb{Q})$, we require a random variable L_T satisfying the following:

- L_T is \mathcal{F}_T measurable, which means that L_T will be known at time T .
- L_T is positive.
- $\mathbb{E}^{\mathbb{P}}[L_T] = 1$.

Define a stochastic process L_t as follows:

$$L_t = \mathbb{E}^{\mathbb{P}}[L_T \mid \mathcal{F}_t], 0 \leq t \leq T$$

The above process is called **Radon-Nikodym derivative**/likelihood process.

Theorem 2.5.

Suppose $X_t, 0 \leq t \leq T$, is an adapted process on Ω, \mathcal{F}_t . We have

$$\mathbb{E}^{\mathbb{Q}}[X_t] = \mathbb{E}^{\mathbb{P}}[L_t \cdot X_t], 0 \leq t \leq T$$

Theorem 2.6 (Bayes' Formula).

For $0 \leq s \leq t \leq T$, we have

$$\mathbb{E}^{\mathbb{Q}}[X_t \mid \mathcal{F}_s] = \frac{1}{L_s} \mathbb{E}^{\mathbb{P}}[L_t \cdot X_t \mid \mathcal{F}_s]$$

3 Martingale & Girsanov

Definition 3.1 (Martingale).

Let $(\Omega, \{\mathcal{F}_t\}_{0 \leq t \leq T}, \mathbb{P})$ be a filtered probability space. Consider an adapted stochastic process $I_t, 0 \leq t \leq T$.

We say that I_t is a **\mathbb{P} -martingale** if

$$\mathbb{E}^{\mathbb{P}}[I_t \mid \mathcal{F}_s] = I_s \text{ for all } 0 \leq s \leq t \leq T$$

Heuristically, we have

$$\mathbb{E}_t[dI_t] = 0$$

Theorem 3.1.

Suppose X_t is an adapted process, and W_t a Brownian motion under \mathbb{P} . Define

$$I_t = \int_0^t X_u dW_u$$

or equivalently,

$$dI_t = X_t dW_t$$

Then I_t is a \mathbb{P} -martingale.

Theorem 3.2 (Martingale Representation Theorem).

Suppose I_t is a \mathbb{P} -martingale, and W_t a Brownian motion under \mathbb{P} . Then there is an adapted process X_t under \mathbb{P} , such that

$$I_t = I_0 + \int_0^t X_u dW_u$$

i.e.,

$$dI_t = X_t dW_t$$

Theorem 3.3 (Girsanov).

Suppose W_t is a Brownian motion under measure \mathbb{P} , and θ a constant. Define

$$\tilde{W}_t = W_t + \theta t$$

i.e.,

$$d\tilde{W}_t = dW_t + \theta dt$$

Then there exists a measure \mathbb{Q} , equivalent to \mathbb{P} , such that \tilde{W}_t is a \mathbb{Q} -Brownian motion.

Moreover, the probability \mathbb{Q} is defined by

$$L_T = \frac{d\mathbb{Q}}{d\mathbb{P}} = e^{-\frac{1}{2}\theta^2 T - \theta W_T}$$

The Radon-Nikodym process L_t in the Girsanov Theorem is given by

$$L_t = E^{\mathbb{P}}[L_T \mid \mathcal{F}_t] = e^{-\frac{1}{2}\theta^2 t - \theta W_t}$$

By Ito's Lemma, L_t has the following equivalent differential form:

$$dL_t = -\theta L_t dW_t$$

Theorem 3.4 (Martinalizing Discounted Stock Price).

In \mathbb{P} , $dS_t = \mu S_t dt + \sigma S_t dW_t$. We denote $\frac{S_t}{M_t}$ the **discounted stock price**, where M_t is the money market account. By Ito's Lemma,

$$d\left(\frac{S_t}{M_t}\right) = \sigma \frac{S_t}{M_t} \left(\frac{\mu - r}{\sigma} dt + dW_t\right)$$

Therefore, we define

$$d\tilde{W}_t = \frac{\mu - r}{\sigma} dt + dW_t$$

to arrive at

$$d\left(\frac{S_t}{M_t}\right) = \sigma \frac{S_t}{M_t} d\tilde{W}_t$$

which makes $\frac{S_t}{M_t}$ a \mathbb{Q} -martingale for a equivalent probability measure $\mathbb{Q} \sim \mathbb{P}$.

Girsanov Theorem ensures \tilde{W}_t is a Brownian motion in \mathbb{Q} , given by the Radon-Nikodym process

$$dL_t = -\theta L_t dW_t$$

where $\theta = \frac{\mu - r}{\sigma}$ is the Sharpe ratio.

The dynamic of S_t in \mathbb{Q} is

$$dS_t = rS_t dt + \sigma S_t d\tilde{W}_t$$

so

$$S_t = S_0 e^{(r - \frac{1}{2}\sigma^2)t + \sigma \tilde{W}_t}$$

3.1 Risk Neutral Valuation

We often denote \mathbb{Q} as the risk neutral measure.

Definition 3.2 (European Contingent Claim).

A **European contingent claim** or a T -claim is a financial instrument consisting of a payment V_T at maturity date T . Here V_T is non-negative random variable.

Theorem 3.5 (Risk-Neutral Valuation Formula).

$$\frac{V_t}{M_t} = E_t^{\mathbb{Q}}\left[\frac{V_T}{M_T}\right]$$

Below is an outline of derivation:

1. Define $U_t := E_t^{\mathbb{Q}}\left[\frac{V_T}{M_T}\right]$. Note U_t is a \mathbb{Q} martingale.
2. Martingale Representation Theorem suggests there exists some adapted process η_t such that

$$dU_t = \eta_t d\tilde{W}_t$$

3. We already have $d\left(\frac{S_t}{M_t}\right) = \sigma \frac{S_t}{M_t} d\tilde{W}_t$. Therefore, $dU_t = \phi_t d\left(\frac{S_t}{M_t}\right)$ where $\phi_t = \frac{\eta_t M_t}{\sigma S_t}$.
4. We define $\Pi_t = \phi_t S_t + \gamma_t M_t$ where $\gamma_t = U_t - \phi_t \frac{S_t}{M_t}$. Therefore, $\Pi_t = U_t M_t$ and $\Pi_T = U_T M_T = V_T$. We claim $U_t M_t$ is arbitrage-free price of derivative.
5. Since Π is self financing, we can show $d\Pi_t = d(U_t M_t) = d(U_t e^{rt})$. Applying Ito's Lemma, $d\Pi_t = rU_t M_t dt + e^{rt} \phi_t d\left(\frac{S_t}{M_t}\right)$. Applying Ito's Lemma one more time and we can arrive at $\Pi_t = \phi_t dS_t + r_t dM_t$, which is the definition of self-financing condition.

6. Then it follows $U_t M_t = V_t$, and theorem follows.

Using this theorem, we can calculate the derivative price at time t from its final payoff at time T , by taking expectation

$$V_t = E^{\mathbb{Q}}[e^{-r(T-t)} V_T \mid \mathcal{F}_t]$$

where one can write $V_T = V_t \cdot f(T-t)$ to get rid of conditional expectation, and take integral eventually after finding out the upper/lower bound of the integral.

Theorem 3.6 (Black's Formula).

Let F_t be the forward contract on risky asset with maturity $T' > 0$. By forward price formula, we have

$$F_t = S_t e^{r(T'-t)} = S_0 e^{rT'} e^{-\frac{1}{2}\sigma^2 t + \sigma \tilde{W}_t} = F_0 e^{-\frac{1}{2}\sigma^2 t + \sigma \tilde{W}_t}$$

Here, F_T is a martingale under \mathbb{Q} since $e^{-\frac{1}{2}\sigma^2 t + \sigma \tilde{W}_t}$ is a L_T term in Girsanov Theorem.

Consider the European call option on this forward contract with option's maturity $T \in (0, T']$ and stike price $K > 0$.

Then the corresponding payoff at maturity T is $G := (F_T - K)^+$. and its price at time 0

$$p_0(G) = E^{\mathbb{Q}}[e^{-rT}(F_T - K)^+]$$

Recognize that $e^{rT}p_0(G)$ corresponds to the BS formula with 0 interest rate. Hence, we can use BS formula directly:

$$e^{rT}p_0(G) = F_0N(d_+) - KN(d_-)$$

$$\text{with } d_{\pm} = \frac{1}{\sigma\sqrt{T}}[\ln(F_0/K) \pm \frac{1}{2}\sigma^2T]$$

Theorem 3.7 (Dividend Paying Asset).

In order to stay self-financing, dividend needs to be used for reinvestment of the stocks. Therefore, we have $S_t^{(q)} = S_te^{qt}$ for the position of stockholder. Then we use the same idea to reduce the problem to a special case of BS equation.

By no-arbitrage, we have, under \mathbb{Q} $S_t^{(q)}$ is a martingale, so

$$S_t^{(q)} = S_0^{(q)}e^{(r-\sigma^2/2)t+\sigma\tilde{W}_t} = S_0e^{(r-\sigma^2/2)t+\sigma\tilde{W}_t}$$

Applying Ito on above equation, we can show that

$$S_t = S_0e^{(r-q-\sigma^2/2)t+\sigma\tilde{W}_t}$$

and now

$$\begin{aligned} E^{\mathbb{Q}}[e^{-rT}(S_T - K)^+] &= e^{-qT} E^{\mathbb{Q}}[e^{-(r-q)T}(S_T - K)^+] \\ &= e^{-qT}[S_0N(d_1) - \tilde{K}^{(q)}N(d_2)] \end{aligned}$$

where $d_{1,2} = \frac{1}{\sigma\sqrt{T}}p\ln(S_0/\tilde{K}^{(q)}) \pm \frac{1}{2}\sigma^2T$ and $\tilde{K}^{(q)} := Ke^{-(r-q)T}$

In general, we have in the case of dividend paying asset

$$c_t = S_te^{-q(T-t)}N(d_1) - Ke^{-r(T-t)}N(d_2)$$

and

$$p_t = Ke^{-r(T-t)}N(-d_2) - S_te^{-q(T-t)}N(-d_1)$$

where

$$d_{1,2} = \frac{1}{\sigma\sqrt{T-t}}[\ln(S_t/K) + (r - q \pm \frac{1}{2}\sigma^2)(T - t)]$$

4 Change of Numeraire

Theorem 4.1 (Change of Numeraire Formula).

Let N_t be a numeraire. There exists a measure \mathbb{Q}^N such that every European-style derivative maturing on the date T , we have the pricing formula

$$\frac{V_t}{N_t} = E_t^{\mathbb{Q}^N}[\frac{V_T}{N_T}]$$

Furthermore, we define \mathbb{Q}^N by

$$L_T = \frac{d\mathbb{Q}^N}{d\mathbb{Q}} = \frac{N_T}{M_T} \cdot \frac{M_0}{N_0}$$

4.1 Introduction to American Options

The distinctive feature of American options is the **early exercise privilege**, where the holder of the option can exercise it at **any time** prior to the option's expiration date.

If exercised at time $t < T$,

- American call option has payoff $(S_t - K)^+$
- American put option has payoff $(K - S_t)^+$.

In general,

Definition 4.1 (American-style Instrument).

An American-style instrument with maturity date T and **payoff function** $\Lambda(S, t)$ is an option that can be exercised at any time before T , and its payoff if exercised at time $t < T$ is given by

$$\Lambda(S, t)$$

where S is the underlying asset price at time t . If the option is not exercised at all, then its payoff at maturity is $\Lambda(S, T)$.

In most circumstances, the strike price K and maturity T are fixed. We define $C(S, t)$ to be the **price** of an American call at time t with $T - t$ to maturity, strike price K and underlying stock price S ; define $P(S, t)$ to be the **price** of an American put at time t with $T - t$ to maturity, strike price K and underlying stock price S .

In discrete-time model, the price of American options, as well as whether to exercise early can be determined via “backward induction”.

Theorem 4.2 (Model-Free Bound).

We have

$$S \geq C(S, t) \geq c(S, t) \geq \max\{0, e^{-r(T-t)}(F(t, T) - K)\}$$

where $F(t, T)$ is the time- t forward price of the underlying stock for delivery at date T .

Similarly,

$$K \geq P(S, t) \geq p(S, t) \geq \max\{0, e^{-r(T-t)}(K - F(t, T))\}$$

Theorem 4.3 (American Call without Dividends).

Assume that interest rates $r > 0$. If underlying stock pays no dividends, then

$$C(S, t) = c(S, t)$$

This is proved by showing at any t , $C(S, t) > S_t - K$.

Similarly, we have the following result for put options:

Theorem 4.4 (American Put).

Assume that interest rate $r > 0$. For every $t < T$, we have, regardless of dividends,

$$p(S, t) < P(S, t)$$

This is proved by showing a strategy that give higher payoff. One particular strategy can be to exercise at time $\min\{u, T\}$ where $u = \min\{t \geq 0 : S_t \leq K - Ke^{-r(T-t)}\}$ and look at the two cases where $u < T$ or $u \geq T$.

5 American Options

The key questions to answer for the pricing of American options are

- When to exercise
- At $t < T$, at what stock price should we exercise

It is known there are no analytic closed form formula for American option price.

5.1 Optimal Exercise Boundary

Recall, at any time $t \leq T$, American put price admits $P(S, t) \geq (K - S)^+$ whereas american call price admits $C(S, t) \geq (S - K)^+$.

For American put, at each time $t < T$, there exists a value $S_*^P(t)$ for the stock price such that

1. If $S \leq S_*^P(t)$, then early exercise is *optimal*, which gives $P(S, t) = (K - S)^+ = K - S \geq 0$
2. If $S > S_*^P(t)$, then immediate exercise is **not** optimal, and we have strict inequality $P(S, t) > (K - S)^+$.

Similarly, for American call, at each time $t < T$, there exists a value $S_*^C(t)$ for the stock price such that

1. If $S \geq S_*^C(t)$, then early exercise is *optimal*, which gives $C(S, t) = (S - K)^+ = S - K \geq 0$
2. If $S < S_*^C(t)$, then immediate exercise is **not** optimal, and we have strict inequality $C(S, t) > (S - K)^+$.

Therefore, we denote $S_*^P(t)$ the **optimal exercise boundary** for American put and $S_*^C(t)$ the **optimal exercise boundary** for American call. We ignore the superscript if type of option is not known.

Theorem 5.1.

1. $S_*^P(t)$ is an **increasing** function of t on $[0, T]$.
2. $S_*^C(t)$ is an **decreasing** function of t on $[0, T]$.

We can use these boundaries to define exercise region E and holding region H .

For American put, $E = \{(S, t) \in D : S \leq S_*^P(t)\}$, $H = \{(S, t) \in D : S > S_*^P(t)\}$.

For American call, $E = \{(S, t) \in D : S \geq S_*^C(t)\}$, $H = \{(S, t) \in D : S < S_*^C(t)\}$.

Theorem 5.2 (Smooth Pasting Condition).

For American put, we have

$$\frac{\partial P}{\partial S} = -1 \text{ at } S = S_*^P(t)$$

For American call, we have

$$\frac{\partial C}{\partial S} = 1 \text{ at } S = S_*^C(t)$$

5.2 Pricing Formulation of American Options

Using delta-hedging, we note that, for American put options, when $(S, t) \in H$, Black Schole's PDE holds, i.e.

$$\mathcal{L}_{BS} = \frac{\partial V}{\partial t} + r \frac{\partial V}{\partial S} S_t + \frac{1}{2} \sigma^2 S_t^2 \frac{\partial^2 V}{\partial S^2} - rV = 0$$

Otherwise, if $(S, t) \in E$, then $\mathcal{L}_{BS} < 0$. (since $d\Pi_t < r\Pi_t dt$)

Specifically, for American put, we also have

$$P(S, t) - (K - S) \geq 0$$

with equality if and only if $(S, t) \in E$. Therefore, for American put, we have

$$\mathcal{L}_{BS}(P) \cdot (P - (K - S)) = 0$$

where $P(S, T) = (K, S)^+$, and $D = \{(S, t) : S > 0, 0 \leq t < T\}$.

In general,

Theorem 5.3 (Linear Complementarity Problem).

Suppose q is the dividend yield of the underlying asset, and let the payoff function be

$$\psi(S) = \begin{cases} S - K & \text{(American call)} \\ K_S & \text{(American put)} \end{cases}$$

LCP is stated as below:

$$\min\left\{-\frac{\partial V}{\partial t} - \frac{1}{2}\sigma^2 S^2 \frac{\partial^2 V}{\partial S^2} - (r - q)S \frac{\partial V}{\partial S} + rV, V - \psi\right\} = 0$$

and

$$V(S, T) = \psi^+$$

where $D = \{(S, t) : S > 0, 0 \leq t < T\}$.

The LCP requires us to solve for $V(S, t)$ and $S_*(t)$ simultaneously.

Take American put option as an example, we have

1. If $S > S_*^P(t)$, then $\mathcal{L}_{BS}(P) = 0$.
2. $P(S_*^P(t), t) = K - S_*^P(t)$.
3. $\frac{\partial P}{\partial S}(S_*^P(t), t) = -1$.
4. $P(S, T) = (K - S)^+$

American call case is similar.

5.3 Optimal Stopping Time

Equivalently, we want to solve what is the optimal stopping time.

Definition 5.1 (Stopping Time).

A **stopping time** is a random variable $\tau : \Omega \rightarrow \mathbb{R}^+$ such that

$$\{\tau > t\} \in \mathcal{F}_t, t > 0$$

Definition 5.2 (Hitting Times).

The hitting time of level x by stochastic process X_t is defined as

$$\tau_x = \inf\{t \in \mathbb{R}^+ : X_t = x\}$$

Define

$$P(S, t) = \sup_{t \leq \tau \leq T} \mathbb{E}_t^{\mathbb{Q}}[e^{-r(\tau-t)}(K - S_\tau)^+]$$

where the supremum is taken over all possible stopping times, then the above supremum is reached at the optimal stopping time τ_{opt} such that

$$\tau_{opt} = \inf_u \{t \leq u \leq T : P(S_u, u) = K - S_u\}$$

We recognize τ_{opt} is the hitting time of the optimal exercise boundary $S_*(t)$, since

- Before hitting boundary, $e^{-rt}P(S_t, t)$ is martingale, and
- after hitting, $e^{-rt}P(S_t, t)$ is a supermartingale

This is because, by Ito's lemma twice,

$$\begin{aligned} d(e^{-rt}P(S_t, t)) &= e^{-rt} \left(\frac{\partial P}{\partial t} + \frac{1}{2} \sigma^2 S_t^2 \frac{\partial^2 P}{\partial S^2} + r S_t \frac{\partial P}{\partial S} - r P \right) dt \\ &\quad + e^{-rt} \sigma S_t \frac{\partial P}{\partial S} dW_t \end{aligned}$$

where $F = \{S_T \geq K, \min_{0 \leq t \leq T} S_t > B\}$.

5.4 Perpetual American Put Options

Perpetual American put options works like usual American put options, except that there is no maturity date. We recognize the price of perpetual American option $P_\infty(S)$, is independent of time t .

Theorem 5.4 (Price of Perpetual American Put).

$$P_\infty(S) = (K - S_*) \left(\frac{S}{S_*} \right)^{\mu_-}$$

Here, $S_* = \frac{\mu_-}{\mu_- - 1} K$, $\mu_\pm = \frac{-(r - \sigma^2/2) \pm \sqrt{(r - \sigma^2/2)^2 + 2r\sigma^2}}{\sigma^2}$.

To derive, use $\frac{\partial P}{\partial t} = 0$ to turn BS PDE into 2nd order PDE. The additional conditions are

- $P_\infty(S_*) = K - S_*$
- $\frac{\partial P_\infty}{\partial S}(S_*) = -1$
- $P_\infty(S) \rightarrow 0$ as $S \rightarrow \infty$.

6 Barrier Option

Barrier options has the payoff the European option subject to whether a prescribed barrier B has been hit. There are two types of conditions:

- Options are activated upon hitting the barrier (knock-in)
- Options are deactivated upon hitting the barrier (knock-out)

Also, there are two types of barrier:

- Downstream barrier, where $B < S$, the current stock price
- Upstream barrier, where $B > S$

We pay particular attention in studying the **down-and-out** call, which pays, at maturity T

$$\begin{cases} (S_T - K)^+ & \text{if the lower barrier is never hit} \\ 0 & \text{otherwise} \end{cases}$$

We denote the price of the down-and-out call with time to maturity T by $c_{do}(S, B, K, T)$, where S is the underlying stock price today at time 0, B the downstream barrier and K the strike price.

The payoff can be written as such:

$$c_{do}(S_T, B, K, 0) = (S_T - K) \mathbf{1}_F$$

Theorem 6.1 (In-out parity).

Knock-out option + knock-in option = vanilla option.

6.1 PDE Formulation

Prior to knock-out, the option is alive and satisfies BS PDE:

$$\frac{\partial V}{\partial t} + r \frac{\partial V}{\partial S} S_t + \frac{1}{2} \sigma^2 S_t^2 \frac{\partial^2 V}{\partial S^2} = rV$$

The conditions imposed by barrier options enter through **boundary conditions** and **solution domains**:

1. When barrier is hit, option becomes worthless: $V(B, t) = 0$ for all $t \in [0, T]$.
2. Solution domain for downstream barrier is: $\{B < S < \infty\} \times [0, T]$.
3. Terminal condition is

$$V(S, T) = \begin{cases} (S - K)^+ & \text{for call} \\ (K - S)^+ & \text{for put} \end{cases}$$

6.2 Martingale Pricing

By risk-neutral valuation, we have

$$c_{do}(S, B, K, T) = e^{-rT} \mathbb{E}^{\mathbb{Q}}[(S_T - K)\mathbf{1}_F]$$

where $F = \{S_T \geq K, \min_{0 \leq t \leq T} S_t > B\}$. Let $V_t^{(1)} = S_t \mathbf{1}_F$ and $V_t^{(2)} = K \mathbf{1}_F$, then by splitting the payoff and observe $\frac{V_T}{M_T}$ is a martingale under \mathbb{Q} , we have

$$c_{do}(S, B, K, T) = V_0^{(1)} - V_0^{(2)}$$

We can then evaluate $V_0^{(1)}$ under \mathbb{Q}^S and $V_0^{(2)}$ under \mathbb{Q} . By change of numeraire, we have

$$\frac{V_0^{(1)}}{S_0} = \mathbb{E}^{\mathbb{Q}^S}\left[\frac{V_T^{(1)}}{S_T}\right] \quad \frac{V_0^{(2)}}{M_0} = \mathbb{E}^{\mathbb{Q}}\left[\frac{V_T^{(2)}}{M_T}\right]$$

Therefore, $V_0^{(1)} = S_0 \mathbb{Q}^S(F)$ and $V_0^{(2)} = K e^{-rT} \mathbb{Q}(F)$. The problem then reduces to find the two probabilities. Note, the stock price has the following dynamic:

- Under \mathbb{Q}^S : $\frac{dS_t}{S_t} = (r + \sigma^2) dt + \sigma dW_t^S$
- Under \mathbb{Q} : $\frac{dS_t}{S_t} = r dt + \sigma dW_t$

Equivalently, we observe the log-prices is of the form, under \mathbb{P} :

$$d(\ln S_t) = \mu dt + \sigma dW_t$$

$$\text{where } \mu = \begin{cases} r + \frac{\sigma^2}{2} & \text{if } \mathbb{P} = \mathbb{Q}^S \\ r - \frac{\sigma^2}{2} & \text{if } \mathbb{P} = \mathbb{Q} \end{cases}.$$

We transform the F 's representation using log prices: $F = \{\ln S_T - \ln S_0 \geq \ln K - \ln S_0, \min_{0 \leq t \leq T} (\ln S_t - \ln S_0) > \ln B - \ln S_0\}$. Let us denote

- $X_t := \ln S_t - \ln S_0$
- $m_T = \min_{0 \leq t \leq T} X_t$
- $x = \ln K - \ln S_0$.
- $m = \ln B - \ln S_0$.

Then $F = \{X_T \geq x, m_T > m\}$, and $X_t = \mu t + \sigma W_t \sim N(\mu t, \sigma^2 t)$ is the Brownian motion with drift under \mathbb{P} , some probabilistic measure.

For our case of down-and-out options, we have $m < 0$, where x can be larger or smaller than m .

instead of computing $\mathbb{P}(F)$, we study the special case $\mathbb{P}(A)$, defined by

$$A = \{W_T \geq x, m_T \leq m\} \text{ with restriction } \underbrace{m \leq 0}_{\text{auto. satisfied}} , \underbrace{m \leq x}_{\text{assumption}}$$

first. By using the reflection principle

$$\mathbb{E}^{\mathbb{P}}[\mathbf{1}_{Ag}(W_T)] = \mathbb{E}^{\mathbb{P}}[\mathbf{1}_{Bg}(2m - W_T)]$$

and a change of measure, we can make $X_T = \sigma \tilde{W}_T$ a brownian motion without drift to simplify computation, where $\tilde{W}_T = W_T + \frac{\mu}{\sigma} T$ is a brownian motion under $\tilde{\mathbb{P}}$. Specifically, the Radon Nikodym derivative is given by

$$\frac{d\mathbb{P}}{d\tilde{\mathbb{P}}} = \exp\left\{-\frac{\mu^2}{2\sigma^2} T + \frac{\mu}{\sigma} \tilde{W}_T\right\} := g(\tilde{W}_T)$$

Then $A = \{\tilde{W}_T \geq \frac{x}{\sigma}, \tilde{m}_T \leq \frac{m}{\sigma}\}$ and $B = \{\tilde{W}_T \leq \frac{2m}{\sigma} - \frac{x}{\sigma}\}$, where $\frac{m}{\sigma} \leq \frac{x}{\sigma}, \frac{m}{\sigma} \leq 0$.

Direct computation yields $\mathbb{P}(A) = \mathbb{E}^{\tilde{\mathbb{P}}}[\mathbf{1}_B \cdot g(\frac{2m}{\sigma} - \tilde{W}_T)] = e^{\frac{2\mu m}{\sigma^2} N(\frac{2m-x+\mu T}{\sigma\sqrt{T}})}$.

Theorem 6.2.

For $m \leq 0$ and $m \leq x$,

- $\mathbb{P}(X_T \geq x, m_T \leq m) = e^{\frac{2\mu m}{\sigma^2}} N(\frac{2m-x+\mu T}{\sigma\sqrt{T}})$
- $\mathbb{P}(X_T \geq x, m_T > m) = N(\frac{\mu T - x}{\sigma\sqrt{T}}) - e^{\frac{2\mu m}{\sigma^2}} N(\frac{2m-x+\mu T}{\sigma\sqrt{T}})$

Suppose in a particular case $x = m$, then event $\{X_T \geq m, m_T > m\} = \{m_T > m\}$. we have, for $m < 0$,

- $\mathbb{P}(m_T > m) = N(\frac{\mu T - m}{\sigma\sqrt{T}}) - e^{\frac{2\mu m}{\sigma^2}} N(\frac{m+\mu T}{\sigma\sqrt{T}})$
- $\mathbb{P}(m_T < m) = N(\frac{m-\mu T}{\sigma\sqrt{T}}) + e^{\frac{2\mu m}{\sigma^2}} N(\frac{m+\mu T}{\sigma\sqrt{T}})$

In general, define the minimum over the time window $[t, T]$ by $m_{[t,T]} = \min_{t \leq u \leq T} X_u$, where $d \iff X_u = \mu du + \sigma dW_u$ with W_u a \mathbb{P} Brownian motion. For $m \leq 0$,

- $\mathbb{P}(m_{[t,T]} > m) = N(\frac{\mu\tau - m}{\sigma\sqrt{\tau}}) - e^{\frac{2\mu m}{\sigma^2}} N(\frac{m+\mu\tau}{\sigma\sqrt{\tau}})$
- $\mathbb{P}(m_{[t,T]} < m) = N(\frac{m-\mu\tau}{\sigma\sqrt{\tau}}) + e^{\frac{2\mu m}{\sigma^2}} N(\frac{m+\mu\tau}{\sigma\sqrt{\tau}})$

where $\tau = T - t$.

Similarly, define the maximum over the time window $[t, T]$ by $M_{[t,T]} = \max_{t \leq u \leq T} X_u$. For $M \geq 0$,

- $\mathbb{P}(M_{[t,T]} \geq M) = N(\frac{\mu\tau - M}{\sigma\sqrt{\tau}}) + e^{\frac{2\mu M}{\sigma^2}} N(\frac{-M-\mu\tau}{\sigma\sqrt{\tau}})$
- $\mathbb{P}(M_{[t,T]} \leq M) = N(\frac{M-\mu\tau}{\sigma\sqrt{\tau}}) - e^{\frac{2\mu M}{\sigma^2}} N(\frac{-M-\mu\tau}{\sigma\sqrt{\tau}})$

6.3 Barrier Option Price

We collect the following parameters d_1, \dots, d_8 . They exist in pairs.

- $d_1 = \frac{\ln \frac{S}{K} + (r - q + \frac{\sigma^2}{2})T}{\sigma\sqrt{T}}$
- $d_2 = d_1 - \sigma\sqrt{T}$.
- $d_3 = \frac{\ln \frac{B^2}{SK} + (r - q + \frac{\sigma^2}{2})T}{\sigma\sqrt{T}}$
- $d_4 = d_3 - \sigma\sqrt{T}$.

- $d_5 = \frac{\ln \frac{S}{B} + (r - q + \frac{\sigma^2}{2})T}{\sigma\sqrt{T}}$
- $d_6 = d_5 - \sigma\sqrt{T}$.
- $d_7 = \frac{\ln \frac{B}{S} + (r - q + \frac{\sigma^2}{2})T}{\sigma\sqrt{T}}$
- $d_8 = d_7 - \sigma\sqrt{T}$.

To price the barrier option where $B < K$,

$$c_{do}(S, B, K, T) = S_0 \mathbb{Q}^S(F) - Ke^{-rT} \mathbb{Q}(F)$$

where $S_0 \mathbb{Q}^S(F) = S_0 N(d_1) - S_0 (\frac{B}{S_0})^{1 + \frac{2r}{\sigma^2}} N(d_3)$, and $-Ke^{-rT} \mathbb{Q}(F) = -Ke^{-rT} N(d_2) + Ke^{-rT} (\frac{B}{S_0})^{\frac{2r}{\sigma^2} - 1} N(d_4)$. Equivalently,

$$c_{do}(S, B, K, T) = c(S, K, T) - (\frac{B}{S})^{\frac{2r}{\sigma^2} - 1} c(\frac{B^2}{S}, K, T)$$

where $c(S, K, T)$ is the price function of the usual European call option with strike K , spot price S and time to maturity T . For case $B \geq K$,

$$c_{do}(S, B, K, T) = SN(d_5) - Ke^{-rT} N(d_6) - (\frac{B}{S})^{1 + \frac{2r}{\sigma^2}} SN(d_7) + (\frac{B}{S})^{\frac{2r}{\sigma^2} - 1} Ke^{-rT} N(d_8)$$

7 Asian Options

Asian options are options whose payoff depends on some form of averaging of underlying asset price. The average will be taken over $[0, T]$. Here, we only consider **European style** options. The options can be

- Fixed Strike, where $(A_T - K)^+$ is the payoff of call option, and $(K - A_T)^+$ of put.
- Floating Strike, where $(S_T - A_T)^+$ is the payoff of call option, and $(A_T - S_T)^+$ for put.

The type of averaging can be

- $\frac{1}{n} \sum_{i=1}^n S_{t_i}$ for discretely sampled arithmetic
- $\frac{1}{T} \int_0^T S_t dt$ for continuously sampled arithmetic
- $\exp(\frac{1}{n} \sum_{i=1}^n \ln S_{t_i})$ for discretely sampled geometric
- $\exp(\frac{1}{n} \int_0^T \ln S_t dt)$ for continuously sampled geometric

We only consider continuously sampled Asian options.

7.1 Multivariate Ito's Lemma

Consider two Ito process evolving over time:

- $dX_t = a(X_t, Y_t) dt + b(X_t, Y_t) dW_t^X$
- $dY_t = c(X_t, Y_t) dt + d(X_t, Y_t) dW_t^Y$
- $\rho = \text{Corr}[dW_t^X, dW_t^Y]$ where ρ is the correlation between dW^X and dW^Y .

Consider another stochastic process $V_t := V(X_t, Y_t)$. Then we have

$$dV_t = \frac{\partial V}{\partial X}(X_t, Y_t) dX_t + \frac{\partial V}{\partial Y}(X_t, Y_t) dY_t + \frac{1}{2} \frac{\partial^2 V}{\partial X^2}(X_t, Y_t) (dX_t)^2 + \frac{1}{2} \frac{\partial^2 V}{\partial Y^2}(X_t, Y_t) (dY_t)^2 + \frac{\partial^2 V}{\partial X \partial Y}(X_t, Y_t) dX_t dY_t$$

Substituting dX_t and dY_t , we will have the multivariate Ito's lemma. When V is also a function of t , then we need to add an extra term $\frac{\partial V}{\partial t} dt$ into the above equation.

7.2 General PDE Framework

The price function $V(S_t, I_t, t)$ for **path-dependent** options depends on S_t, I_t, t , where I_t is the **path-dependent variable**

$$I_t = \int_0^t f(S_u, u) du$$

Equivalently, $dI_t = f(S_t, t) dt$.

As usual, we assume under \mathbb{Q} , stock price follows $dS_t = rS_t dt + \sigma S_t dW_t$.

Applying multivariate Ito Lemma to V_t , we have

$$dV_t = (\frac{\partial V}{\partial t} + \frac{1}{2} \sigma^2 S_t^2 \frac{\partial^2 V}{\partial S^2}) dt + \frac{\partial V}{\partial S} dS_t + \frac{\partial V}{\partial I} dI_t$$

since $(dI_t)^2 = 0$ and $dS_t dI_t = 0$.

Using same argument of delta hedging, we arrive at

$$\frac{\partial V}{\partial t} + \frac{1}{2} \sigma^2 S^2 \frac{\partial^2 V}{\partial S^2} + rS \frac{\partial V}{\partial S} + f(S, t) \frac{\partial V}{\partial I} - rV = 0$$

Set the continuously sampled arithmetic mean as

$$A_T = \frac{1}{T} \int_0^T S_u du$$

We want to obtain $f(S_t, t)$ that contains information about A_T .

- $f(S_t, t) = S_t$.
Then $I_t = \int_0^t S_u du$, and $dI_t = S_t dt$. This allows us to relate $A_T = \frac{I_T}{T}$.
The PDE is

$$\frac{\partial V}{\partial t} + \frac{1}{2} \sigma^2 S^2 \frac{\partial^2 V}{\partial S^2} + rS \frac{\partial V}{\partial S} + S \frac{\partial V}{\partial I} - rV = 0$$

The terminal condition is

$$V(S, I, T) = \begin{cases} (I/T - S)^+ & \text{floating put} \\ (S - I/T)^+ & \text{floating call} \\ (I/T - K)^+ & \text{fixed strike call} \\ (K - I/T)^+ & \text{fixed strike put} \end{cases}$$

- $f(S_t, t) = \frac{1}{t}(S_t - A_t)$ where $A_t = \frac{1}{t} \int_0^t S_u \, du$. By Ito, $dA_t = \frac{1}{t}(S_t - A_t) \, dt$, or equivalently, $A_t = \int_0^t \frac{1}{u}(S_u - A_u) \, du + S_0$.

The PDE will be of the form

$$\frac{\partial V}{\partial t} + \frac{1}{2}\sigma^2 S^2 \frac{\partial^2 V}{\partial S^2} + rS \frac{\partial V}{\partial S} + \frac{S_A}{t} \frac{\partial V}{\partial A} - rV = 0$$

and the terminal condition is

$$V(S, A, T) = \begin{cases} (A - S)^+ & \text{floating put} \\ (S - A)^+ & \text{floating call} \\ (A - K)^+ & \text{fixed strike call} \\ (K - A)^+ & \text{fixed strike put} \end{cases}$$

7.3 Roger-Shi Method

We investigate the second approach. We define the new variable

$$X_t = \frac{K - \frac{1}{T} \int_0^t S_u \, du}{S_t} = \frac{K - \frac{I_t}{T}}{S_t}$$

with $X_0 = \frac{K}{S_0}$.

This allows $V(S, I, t) = e^{-r(T-t)} \mathbb{E}_t^{\mathbb{Q}}[(A_T - K)^+]$.

Note,

$$\mathbb{E}_t^{\mathbb{Q}}[(A_T - K)^+] = S_t \mathbb{E}_t^{\mathbb{Q}}[(\frac{1}{T} \int_t^T \frac{S_u}{S_t} \, du - X_t)^+]$$

Define $h(X, t) = \mathbb{E}_t^{\mathbb{Q}}[(\frac{1}{T} \int_t^T \frac{S_u}{S_t} \, du - X)^+]$ and let

$$H(X, t) = e^{-r(T-t)} h(X, t)$$

Then we have

$$V(S, I, t) = SH(X, t)$$

The resulting PDE with respect to H is

$$\frac{\partial H}{\partial t} + \frac{1}{2}\sigma^2 X^2 \frac{\partial^2 H}{\partial X^2} - (\frac{1}{T} + rX) \frac{\partial H}{\partial X} = 0$$

where $-\infty < X < \infty$ and $t \in [0, T]$, with terminal condition $V(S, I, T) = (\frac{I}{T} - K)^+$, or equivalently $H(X, T) = (-X)^+$.

The closed form analytic solution exists to $H(X, t)$ if $X \leq 0$. This is because the terminal condition is simplified to $H(X, T) = -X \geq 0$, a linear equation.

Analytic solution of $H(X, t)$ is

$$H(X, t) = \frac{1 - e^{-r(T-t)}}{rT} - e^{-r(T-t)} X$$

and

$$V(S, I, t) = (\frac{I}{T} - K)e^{-r(T-t)} + \frac{1 - e^{-r(T-t)}}{rT} S$$

However, there is no closed form analytic solution for $X > 0$. One need to solve this equation:

$$\frac{\partial H}{\partial t} + \frac{1}{2}\sigma^2 X^2 \frac{\partial^2 H}{\partial X^2} - (\frac{1}{T} + rX) \frac{\partial H}{\partial X} = 0$$

with domain $t \in [0, T]$, $x \in (0, \infty)$ and terminal condition $H(0, t) = \frac{1 - e^{-r(T-t)}}{rT}$, $H(\infty, t) = 0$ and $H(X, T) = -X$, $X \in (0, \infty)$.

7.4 Put call Parity for Fixed Strike Arithmetic Asian Option

The put call parity is

$$\begin{aligned} c_{\text{fix}}(S_t, I_t, T) - p_{\text{fix}}(S_t, I_t, T) \\ = (\frac{I_t}{T} - K)e^{-r(T-t)} + \frac{1 - e^{-r(T-t)}}{rT} S_t \end{aligned}$$

7.5 Geometric Asian option

Geometric Asian option in continuous manner has the running geometric $G_t = \exp(\frac{1}{t} \int_0^t \ln S_u \, du)$. By Ito lemma,

$$dG_t = \frac{G_t \ln \frac{S_t}{G_t}}{t} \, dt$$

The PDE is

$$\frac{\partial V}{\partial t} + \frac{1}{2}\sigma^2 S^2 \frac{\partial^2 V}{\partial S^2} + rS \frac{\partial V}{\partial S} + \frac{G \ln \frac{S}{G}}{t} \frac{\partial V}{\partial G} - rV = 0$$

The solution domain is $\{S > 0, G > 0, t \in [0, T]\}$.

Let's try to solve for the fixed strike geometric Asian. Denote the time- t price by $c_{\text{fix}}(S, G, t)$ where $0 \leq t \leq T$. Risk neutral valuation asserts

$$c_{\text{fix}}(S, G, t) = e^{-r(T-t)} \mathbb{E}_t^{\mathbb{Q}}[(G_T - K)^+]$$

For $t \leq T$, define $X_t = \frac{1}{t} \int_0^t \ln S_u \, du$. This gives $G_t = e^{X_t}$, and the payoff is $(e^{X_T} - K)^+$.

By studying the distribution of $\ln S_u$, we have

$$\begin{aligned} X_T \\ &= \frac{t}{T} X_t + \frac{1}{T} \int_t^T (\ln S_t + (r - \frac{\sigma^2}{2})(u - t) + \sigma(W_u - W_t)) \, du \\ &= \frac{t}{T} X_t + \frac{T-t}{T} \ln S_t \\ &\quad + (r - \sigma^2/2) \frac{(T-t)^2}{2T} + \frac{\sigma}{T} (\int_t^T (W_u - W_t) \, du) \end{aligned}$$

Note, the last term is normally distributed with mean 0 and variance $\frac{\sigma^2}{T^2} \frac{(T-t)^3}{3}$.

By setting

- $\bar{\mu} = (r - \sigma^2/2) \frac{(T-t)^2}{2T}$
- $\bar{\sigma} = \frac{\sigma}{T} \sqrt{\frac{(T-t)^3}{3}}$

Therefore, we have

$$G_T = G_t^{t/T} S_t^{(T-t)/T} \exp(\bar{\mu} + \bar{\sigma} \phi)$$

Using normal calculation, we have

$$c_{\text{fix}}(S, G, t) = e^{-r(T-t)} (G_t^{t/T} S_t^{(T-t)/T} e^{\bar{\mu} + \bar{\sigma}^2/2} N(d_1) - K N(d_2))$$

where $d_2 = \frac{1}{\bar{\sigma}} (\frac{t}{T} \ln G_t + \frac{T-t}{T} \ln S_t - \ln K + \bar{\mu})$ and $d_1 = d_2 + \bar{\sigma}$.

8 Lookback Options & Two Assets Options

8.1 Lookback Option

Lookback options are path dependent options whose payoffs depends on the **maximum** or **minimum** of the underlying stock price attained over a certain period of time, known as **lookback period**. We set the lookback period to be $[0, T]$.

Let minimum value of underlying asset over the lookback period $[0, T]$ by

$$m_T = \min_{0 \leq t \leq T} S_t$$

and the maximum value by

$$M_T = \max_{0 \leq t \leq T} S_t$$

There are 2 types of lookback options:

1. Fixed Strike:

- Call: $(M_T - K)^+$
- Put: $(K - m_T)^+$

2. Floating Strike:

- Call: $(S_T - m_T)$
- Put: $(M_T - S_T)$

Recall, the general PDE for European path-depended options is

$$\frac{\partial V}{\partial t} + \frac{1}{2} \sigma^2 S^2 \frac{\partial^2 V}{\partial S^2} + r S \frac{\partial V}{\partial S} + f(S, t) \frac{\partial V}{\partial I} - r V = 0$$

where $I_t = \int_0^t f(S_u, u) du$ is the path dependent variable.

Theorem 8.1. Suppose $f(x)$ continuous and positive on $[a, b]$. The maximum of $f(x)$ on $[a, b]$ can be given by $\max_{a \leq x \leq b} f(x) = \lim_{n \rightarrow \infty} (\int_a^b f(x)^n dx)^{\frac{1}{n}}$.

Theorem 8.2. It follows that $M_t = \lim_{n \rightarrow \infty} (\int_0^t S_u^n du)^{\frac{1}{n}}$ and $m_t = \lim_{n \rightarrow \infty} (\int_0^t S_u^{-n} du)^{-\frac{1}{n}}$.

Define $I_{n,t} = \int_0^t S_u^n du$, and $M_{n,t} = I_{n,t}^{\frac{1}{n}}$. Then $M_t = \lim_{n \rightarrow \infty} M_{n,t}$. By Ito's Lemma,

$$d I_{n,t} = S_t^n dt$$

and

$$d M_{n,t} = \frac{1}{n} \frac{S_t^n}{M_{n,t}^{n-1}} dt$$

Ito's Lemma gives

$$\begin{aligned} d V_t &= \frac{\partial V}{\partial t} dt + \frac{\partial V}{\partial S} d S + \frac{\partial V}{\partial M_n} d M_{n,t} + \frac{1}{2} \frac{\partial^2 V}{\partial S^2} (d S_t)^2 \\ &= \left(\frac{\partial V}{\partial t} + \frac{1}{2} \sigma^2 S_t^2 \frac{\partial^2 V}{\partial S^2} + \frac{1}{n} \frac{S_t^n}{M_{n,t}^{n-1}} \frac{\partial V}{\partial M_n} \right) dt + \frac{\partial V}{\partial S} d S_t \end{aligned}$$

Note, $(d(M_{n,t}))^2$ does not appear since $d(M_{n,t})$ is deterministic. Then by Delta hedging and self financing, we arrive at

$$d \Pi_t = \left(\frac{\partial V}{\partial t} + \frac{1}{2} \sigma^2 S_t^2 \frac{\partial^2 V}{\partial S^2} \right) dt$$

The term $\frac{1}{n} \frac{S_t^n}{M_{n,t}^{n-1}} \leq \frac{S_t}{n} \rightarrow 0$ as $n \rightarrow \infty$. This gives back the Black-Scholes PDE, where

$$\frac{\partial V}{\partial t} + r \frac{\partial V}{\partial S} S_t + \frac{1}{2} \sigma^2 S_t^2 \frac{\partial^2 V}{\partial S^2} = r V$$

For American lookback options, we have $\mathcal{L}_{BS} \leq 0$ with equality if and only if (S, t) is in holding region.

For European floating strike lookback put option, it satisfies BS PDE with following boundary and terminal conditions

1. $V(S, M, T) = M - S$
2. $V(0, M, t) = e^{-r(T-t)} M$
3. $\frac{\partial V}{\partial M}(M, M, t) = 0$

Solution domain is $\{(S, M, t) : 0 < S < M, 0 \leq t \leq T\}$.

8.2 Two Asset Options

Suppose $d X_t = \mu_X X_t dt + \sigma_X X_t d W_t^X$ and $d Y_t = \mu_Y Y_t dt + \sigma_Y Y_t d W_t^Y$ are two stocks with $d W_t^X d W_t^Y = \rho dt$. We can use Delta-hedging to derive the PDE. The key point is Multivariate Ito's Lemma which gives

$$\begin{aligned} d V_t &= \left(\frac{\partial V}{\partial t} + \frac{1}{2} \sigma_X^2 X_t^2 \frac{\partial^2 V}{\partial X^2} + \frac{1}{2} \sigma_Y^2 Y_t^2 \frac{\partial^2 V}{\partial Y^2} \right. \\ &\quad \left. + \rho \sigma_X \sigma_Y X_t Y_t \frac{\partial^2 V}{\partial X \partial Y} \right) dt \\ &\quad + \frac{\partial V}{\partial X} d X_t + \frac{\partial V}{\partial Y} d Y_t \end{aligned}$$

Therefore, $\Pi_t = V_t - \frac{\partial V}{\partial X}X_t - \frac{\partial V}{\partial Y}Y_t$ in order to be riskless. Finally, we have

$$\begin{aligned} \frac{\partial V}{\partial t} + \frac{1}{2}\sigma_X^2 X^2 \frac{\partial^2 V}{\partial X^2} + \frac{1}{2}\sigma_Y^2 Y^2 \frac{\partial^2 V}{\partial Y^2} + \rho\sigma_X\sigma_Y XY \frac{\partial^2 V}{\partial X\partial Y} \\ + rX \frac{\partial V}{\partial X} + rY \frac{\partial V}{\partial Y} - rV = 0 \end{aligned}$$

The solution domain is $\{X > 0, Y > 0, t \in [0, T]\}$, with terminal condition $V(X, Y, T) = f(X, Y)$.

8.3 Exchange Options

European Exchange Options that allows exchange of Y_T for X_T has the terminal payoff at maturity

$$V(X_T, Y_T, T) = \max\{X_T - Y_T, 0\}$$

It obeys the two asset option PDE. However, to make it solvable, we let $Z = \frac{X}{Y}$. So that $V(X, Y, t) = Y \cdot H(Z, t)$, and we solve for H .

The partial derivatives are

- $\frac{\partial V}{\partial t} = Y \frac{\partial H}{\partial t}$
- $\frac{\partial V}{\partial Y} = H - Z \frac{\partial H}{\partial Z}$
- $\frac{\partial^2 V}{\partial Y^2} = \frac{1}{Y} Z^2 \frac{\partial^2 H}{\partial Z^2}$
- $\frac{\partial^2 V}{\partial X \partial Y} = -\frac{X}{Y^2} \frac{\partial^2 H}{\partial Z^2}$.

Therefore, we arrive at

$$\frac{\partial H}{\partial t} + \frac{1}{2}\bar{\sigma}^2 Z^2 \frac{\partial^2 H}{\partial Z^2} = 0$$

where $\bar{\sigma}^2 = \sigma_X^2 + \sigma_Y^2 - 2\rho\sigma_X\sigma_Y$, and terminal condition for H is $H(Z, T) = (Z - 1)^+$. This can be solved as

$$H(Z, t) = ZN(d_1) - N(d_2)$$

where $d_1 = \frac{\ln Z + \frac{\bar{\sigma}^2}{2}(T-t)}{\bar{\sigma}\sqrt{T-t}}$, $d_2 = d_1 - \bar{\sigma}\sqrt{T-t}$. Consiquently, $V(X, Y, t) = XN(d_1) - YN(d_2)$.

8.4 Cross-currency Options

Suppose underlying aaset S_t is in foreign currency f and payoff in domestic currency d . The payoff is affected by terminal asset price S_T in foreign currency and terminal exchange rate F_T .

Definition 8.1. The exchange rate F_t is the price of one unit foreign currency in US dollars(domestic).

The model is $dS_t = \mu_S S_t dt + \sigma_S S_t dW_t^S$ and $dF_t = \mu_F F_t dt + \sigma_F F_t dW_t^F$ with $dW_t^S dW_t^F = \rho dt$.

Let $V(S, F, t)$ define the options value in domestic currency at time t . We use delta-hedging:

$$\Pi_t = V_t - \delta_F \cdot F_t - \delta_S \cdot S_t \cdot F_t$$

Let r_f be forieng risk-free rate. This can be treated as dividend yield of foriegn currency.

Self financing condition gives

$$\begin{aligned} d\Pi_t = dV - \Delta_F dF - \underbrace{\Delta_F r_f F dt}_{\text{dividend}} \\ - \Delta_S \underbrace{(S dF + F dS + \rho\sigma_S\sigma_F SF dt)}_{d(SF)} \end{aligned}$$

where dV is given by

$$\begin{aligned} dV = \left(\frac{\partial V}{\partial t} + \frac{1}{2}\sigma_S^2 S^2 \frac{\partial^2 V}{\partial S^2} + \frac{1}{2}\sigma_F^2 F^2 \frac{\partial^2 V}{\partial F^2} \right. \\ \left. + \rho\sigma_S\sigma_F SF \frac{\partial^2 V}{\partial F \partial S} \right) dt \\ + \frac{\partial V}{\partial S} dS + \frac{\partial V}{\partial F} dF \end{aligned}$$

This gives $\Delta_S = \frac{1}{F} \frac{\partial V}{\partial S}$ and $\Delta_F = \frac{\partial V}{\partial F} - \frac{S}{F} \frac{\partial V}{\partial S}$. Furthermore, $d\Pi = r_d \Pi dt$. Simplification gives

$$\begin{aligned} \frac{\partial V}{\partial t} + \frac{1}{2}\sigma_S^2 S^2 \frac{\partial^2 V}{\partial S^2} + \frac{1}{2}\sigma_F^2 F^2 \frac{\partial^2 V}{\partial F^2} \\ + \rho\sigma_S\sigma_F SF \frac{\partial^2 V}{\partial F \partial S} + (r_d - r_f)F \frac{\partial V}{\partial F} + (r_f - \rho\sigma_S\sigma_F)S \frac{\partial V}{\partial S} \\ - r_d V = 0 \end{aligned}$$