

Theoretical Grounds and Market Adaptations of Financial Fx and Interest Rate Options

Gerardo Durán Martín

Universidad Marista

August 23, 2017



Table of Contents

1 Financial Markets

2 Probability and Measure Theory

3 Option Pricing: A primer



Some Important Definitions

Definition (Security)

A security is an instrument that either represents ownership or that derives its value from a commodity.

Some Important Definitions

Definition (Security)

A security is an instrument that either represents ownership or that derives its value from a commodity.

Definition (Market)

A market is the place where buyers meet sellers to exchange securities.

Some Important Definitions

Definition (Security)

A security is an instrument that either represents ownership or that derives its value from a commodity.

Definition (Market)

A market is the place where buyers meet sellers to exchange securities.

Definition (Arbitrage)

We define an arbitrage as the probability of making a risk-free profit from a suitable market strategy.

$$\mathbb{P}(\text{Risk-free profit} > 0) = 1$$

Gerardo Durán Martín

Market Assumptions

- Many buyers and sellers (liquidity);
- possible to lend and borrow at the same rate interest rate r ;
- Market participants take advantage of arbitrage opportunity, thereby correcting any deviation from the actual value of any asset.
- All participants have the same information.

- Bonds

A Bond is a debt obligation, its main function is to raise capital for the issuer of the bond. In turn, the buyer of the bond receives interest on the amount loaned.

The Instruments

- Bonds
- Stocks

A stock is a security that represents ownership on a fraction of a corporation. The return on the company for the owner of a stock is represented as a dividend.

The Instruments

- Bonds
- Stocks
- Foreign Exchange Currencies

“One countrys currency freely convertible in the foreign exchange market.” (Kozikowski, 2013)

The Instruments

- Bonds
- Stocks
- Foreign Exchange Currencies
- **Derivatives**

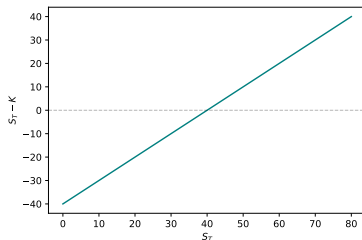
“[A derivative is] a financial instrument whose value depends on (or derives from) the values of other, more basic, underlying variables.”
(Hull, 2014)

Derivatives: An Example

Definition (Forward)

A forward is a derivative contract that gives the buyer both the right, and the obligation to purchase a specified amount of the stock at some future time T , at a price K . The value of the forward today is 0.

The payoff of the forward is $S_T - K$. What is the K such that the contract has zero value today and has no possibility of arbitrage?



Derivatives: An Example (cont'd)

Assume a continuously compounded interest rate r , denote S_t the value of the stock at time t . At $t = T$ the value of the forward is

$$S_T - K = 0$$

By no arbitrage, the present value of the strategy is

$$S_0 - Ke^{-rT} = 0 \implies K = S_0 e^{rT}$$

Therefore, $S_0 e^{rT}$ is the the value that guarantees no arbitrage.



Derivatives: An Example (cont'd)

(Why?)

Assume $K' > K = S_0 e^{rT}$.

- At $t = 0$, borrow S_0 ; purchase the stock; and take a short position the forward contract.
- At $t = T$, receive K' ; give the forward; and repay $S_0 e^{rT}$. Profit $K' - S_0 e^{rT}$.

The same can be said for $K' < K$, by taking a long position.

Derivatives: Another Example

Definition (European Call Option)

A European call option is a derivative contract that gives the buyer the right, but not the obligation to purchase a specified amount of stock at some future time T , at a price K .

The payoff of the option is $\max\{S_T - K, 0\}$. What is the price of the option today that guarantees no arbitrage?

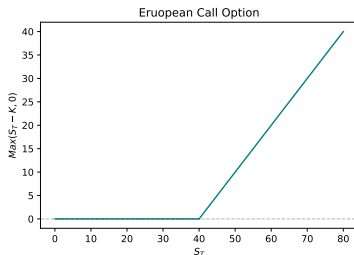


Table of Contents

1 Financial Markets

2 Probability and Measure Theory

3 Option Pricing: A primer



We will work on the following probability space.

Definition (Probability Space: Kolmogorov's Axiom System)

An ordered tripe $(\Omega, \mathcal{F}, \mathbb{P})$ where

- Ω is a set of points ω ;
- \mathcal{F} is a σ -algebra of elements of Ω ; and
- \mathbb{P} is a probability on \mathcal{F} .

Definition (Random Variable)

$$X^{-1}(B) = \{\omega \in \Omega | X(\omega) \in B\} \quad \forall B \subseteq \mathcal{B}(\mathbb{R}).$$

Definition (Integral v.1)

Let $(\Omega, \mathcal{F}, \mu)$ be a measure space, we define the Lebesgue integral (or integral) of the indicator function $\mathbb{1}_A$ w.r.t. μ as

$$\int_{\Omega} \mathbb{1}_A d\mu := \mu(A)$$

Definition (Simple Function)

Let (Ω, \mathcal{F}) a measurable space and X an \mathcal{F} -measurable function. It is said that X is a **simple function** if it takes only a unique finite number of values $\{x_i\}_{i=1}^n \in \mathbb{R}$ over measurable sets $\{A_i\}$. Then, X can be written as

$$X(\omega) = \sum_{i=1}^n x_i \mathbb{1}_{A_i}$$

Definition (Integral v.2)

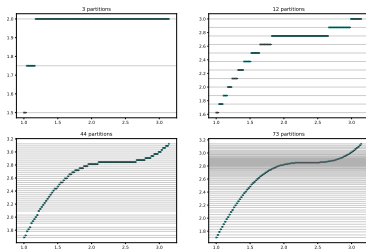
let X be a simple function, we define the integral of X w.r.t. μ as:

$$\int_{\Omega} X(\omega) d\mu(\omega) := \sum_{i=1}^n x_i \mu(A_i)$$

Definition (Integral v.3)

For any nonnegative function X , we define the integral of X w.r.t. a measure μ as

$$\int_{\Omega} X d\mu := \sup \left\{ \int_{\Omega} h d\mu \mid 0 \leq h \leq X, h \text{ is simple} \right\}$$



If X is any measurable function, we define the positive and negative parts of X as:

$$X^+ := \max\{X(\omega), 0\} \quad (1)$$

$$X^- := \max\{-X(\omega), 0\} \quad (2)$$

Thus, $X = X^+ - X^-$, and $|X| = X^+ + X^-$. Since both parts of X are positive, their integrals are well defined.

Definition (L_1 spaces)

We denote by $L_1 \equiv L_1(\Omega, \mathcal{F}, \mu)$ the family of integrable functions w.r.t. μ . Note that $X \in L_1$ if and only if $|X| \in L_1$. i.e.,

$$\int_{\Omega} |X| d\mu = \int_{\Omega} X^+ d\mu + \int_{\Omega} X^- d\mu < \infty \quad (3)$$

Proposition

For any X, Y \mathcal{F} -measurable functions in L_1 ; A, B members of \mathcal{F} :

$$X \geq 0 \implies \int_{\Omega} X d\mu \geq 0$$

If $X \leq Y$ (monotonicity),

$$\int_{\Omega} X d\mu \leq \int_{\Omega} Y d\mu$$

If $A \subseteq B$ and $X \geq 0$,

$$\int_A X d\mu \leq \int_B X d\mu$$

If $\alpha, \beta \in \mathbb{R}$ (linearity),

$$\int_{\Omega} (\alpha X + \beta Y) d\mu = \alpha \int_{\Omega} X d\mu + \beta \int_{\Omega} Y d\mu$$

Theorem (Radon Nikodym Theorem)

Let μ and λ be σ -finite positive measures defined on (Ω, \mathcal{F}) such that, for every $A \in \mathcal{F}$, $\lambda(A) = 0 \implies \mu(A) = 0$. Then, there exists a function $f : \Omega \rightarrow [0, \infty]$ such that

$$\mu(A) = \int_A f d\lambda.$$

Where the function f is defined up to sets with measure zero. f is sometimes called the Radon-Nikodym derivative and it can be written as $\frac{d\mu}{d\lambda}$

Expectation

Definition

Let X be a r.v. on $(\Omega, \mathcal{F}, \mathbb{P})$, we define the expectation of X with respect to (w.r.t.) \mathbb{P} as

$$\mathbb{E}[X] := \int_{\Omega} X(\omega) d\mathbb{P}(\omega)$$

Definition (Conditional Expectation w.r.t. a σ -algebra)

The conditional expectation of a nonnegative random variable X with respect to the σ -algebra \mathcal{G} is a nonnegative random variable denoted $\mathbb{E}[X|\mathcal{G}]$ or $\mathbb{E}[X|\mathcal{G}](\omega)$ such that,

- ① $\mathbb{E}[X|\mathcal{G}]$ is \mathcal{G} -measurable; and
- ② For every $A \in \mathcal{G}$,

$$\int_A X d\mathbb{P} = \int_A \mathbb{E}[X|\mathcal{G}] d\mathbb{P}$$

Proposition

- ① If $a, b \in \mathbb{R}$, $\mathbb{E}[\alpha X + \beta Y | \mathcal{G}] = \alpha \mathbb{E}[X | \mathcal{G}] + \beta \mathbb{E}[Y | \mathcal{G}]$;
- ② $\mathbb{E}[\mathbb{E}[X | \mathcal{F}]] = \mathbb{E}[X]$;
- ③ $\mathbb{E}[\mathbb{E}[X | \mathcal{F}_2] | \mathcal{F}_1] = \mathbb{E}[X | \mathcal{F}_1]$ if $\mathcal{F}_1 \subseteq \mathcal{F}_2$;
- ④ $\mathbb{E}[\mathbb{E}[X | \mathcal{F}_2] | \mathcal{F}_1] = \mathbb{E}[X | \mathcal{F}_2]$ if $\mathcal{F}_2 \subseteq \mathcal{F}_1$; and
- ⑤ Let X be a random variable such that $X \perp \mathcal{G}$ then, $\mathbb{E}[X | \mathcal{G}] = \mathbb{E}[X]$.
Consequently, for any borel-measurable function h ,
 $\mathbb{E}[h(X) | \mathcal{G}] = \mathbb{E}[h(X)]$.

Definition (Stochastic Process)

Let $(\Omega, \mathcal{F}, \mathbb{P})$ be a probability space. A stochastic process is a set of random variables that takes values in some set S called the **state space**, and are indexed by some set T .

$$\{X_t \mid t \in T\}. \quad (4)$$

We can consider a stochastic process as a two variable function

$$X : T \times \Omega \rightarrow S$$

Where,

- 1 For fixed $t \in T$, $\omega \rightarrow X_t(\omega)$ is random variable; and
- 2 for fixed $\omega \in \Omega$, $t \rightarrow X_t(\omega)$ is the path of the process.

Stochastic Processes

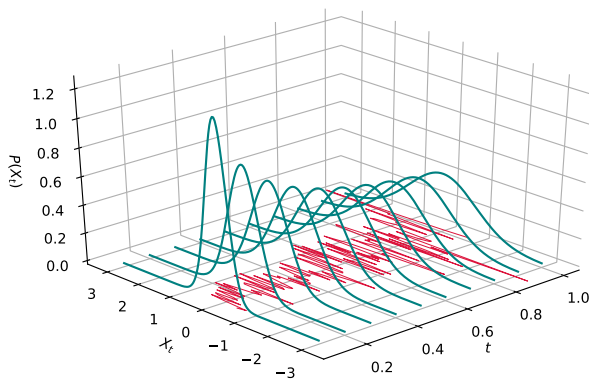


Figure: Distribution and Path of a Stochastic Process.

Definition (Martingale)

An stochastic process $\{X_t\}_{t \geq 0}$ defined over $(\Omega, \mathcal{F}, \mathbb{P})$ is said to be a martingale w.r.t. a filtration $\{\mathcal{F}_n\}_{n \geq 0}$ if, for all $t \geq 0$, and $t \leq s$,

- 1 X_t is \mathcal{F}_t -measurable;
- 2 X_t is in L_1 (i.e. is measurable for all t); and
- 3 $\mathbb{E}[X_t | \mathcal{F}_s] = X_t$.

Table of Contents

1 Financial Markets

2 Probability and Measure Theory

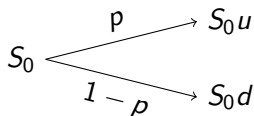
3 Option Pricing: A primer



One Step Binomial Model

We assume a simple market consisting of a stock and a cash bond.

- **The Stock**

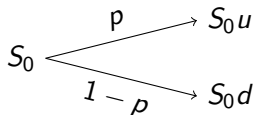


Stock has value S_0 at outset. Between any two units of time, the stock can either go up by a factor u (with probability p), or down by a factor d (with probability $1 - p$).

One Step Binomial Model

We assume a simple market consisting of a stock and a cash bond.

- **The Stock**

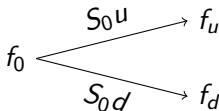


Stock has value S_0 at outset. Between any two units of time, the stock can either go up by a factor u (with probability p), or down by a factor d (with probability $1 - p$).

- **The Cash Bond**

Represents the time value of money. B_0 invested at $t = 0$ grows to $B_0 e^{rT}$ one period thereafter.

This simple market yields the possibility for a derivatives' market, i.e., there could be a payoff with price f_0 today that derives its value from the possible directions the stock may take: either f_u if it goes up, or f_d otherwise.



(Why?) Consider the portfolio (ϕ, ψ) with ϕ number of stocks, and ψ amount in the bank. The value of the portfolio at $t = 0$ is

$$\phi S_0 + \psi B_0$$

We derive a payoff f contingent of the movement of the stock. At $t = \delta_t$ we have

$$\phi S_0 u + \psi B_0 e^{r\delta_t} = f_u$$

$$\phi S_0 d + \psi B_0 e^{r\delta_t} = f_d$$

Yields,

$$\phi = \frac{f_u - f_d}{S_0(u - d)} \quad (4)$$

$$\psi = e^{-r\delta_t} \frac{uf_d - df_u}{u - d}$$

Therefore, buying the portfolio (ϕ, ψ) guarantees the payoff at $t = \delta_t$ (a risk-free strategy). Denote \mathcal{V} the value of the portfolio at $t = 0$.

$$\mathcal{V} = e^{-r\delta_t} \left[f_u \frac{e^{r\delta_t} - d}{u - d} + f_d \frac{u - e^{r\delta_t}}{u - d} \right]. \quad (6)$$

\mathcal{V} is the cost of a risk-free strategy that guarantees the payoff. Furthermore, showing that $d < \exp(r\delta_t) < u$ proves that there exists no possibility of arbitrage.

Denote

$$q := \frac{\exp(r\delta_t) - d}{u - d} \implies 1 - q = \frac{u - \exp(r\delta_t)}{u - d}. \quad (7)$$

We rewrite (6) as

$$\mathcal{V} = \exp(-r\delta_t)[S_0 u \cdot q + S_0 d \cdot (1 - q)] = \mathbb{E}_{\mathbb{Q}}[\mathcal{V}_{\delta_t} | \mathcal{F}_0]. \quad (8)$$

\mathcal{V} is the discounted expectation under a probability measure \mathbb{Q} , and a martingale.

- [1] Dave Applebaum. Lecture notes in measure and probability, 2016.
- [2] Jakša Cvitanic and Fernando Zapatero. *Introduction to the Economics and Mathematics of Financial Markets*. MIT Press, 2004. ISBN 0262033208.
- [3] Larry Harris. *Trading and Exchanges: Market Microstructure for Practitioners*. Oxford University Press, 2002. ISBN 0195144708.
- [4] Onésimo Hernández-Lerma. Probabilidad y procesos estocásticos, September 2009.
- [5] John C. Hull. *Options Futures and Other Derivatives*. Prentice Hall/Pearson, 2014. ISBN 0133456315.
- [6] Zbigniew Kozikowski. *Finanzas Internacionales*. McGraw Hill, 2013. ISBN 6071509777.
- [7] Office of Investor Education and Advocacy. What are corporate bonds? URL https://www.sec.gov/investor/alerts/ib_corporatebonds.pdf. SEC Pub. No. 149.
- [8] Bernt Øksendal. *Stochastic Differential Equations: An Introduction with Applications*. Springer-Verlag Berlin Heidelberg, 2003. ISBN 9783642143946.

- [9] Luis Rincón. *Curso Intermedio de Probabilidad*. Facultad de Ciencias UNAM, 2007.
- [10] Albert N. Shiryaev. *Probability-1*. Springer, 2016. ISBN 038772205X.
- [11] Steven Shreve. *Stochastic Calculus for Finance II*. Springer-Verlag New York, 2004. ISBN 9780387401010.
- [12] Comisión Nacional Bancaria y de Valores. *Glosario de Términos Portafolio de Información*. URL http://portafolioinfo.doctos.cnbv.gob.mx/Documentacion/minfo/00_DOC_R1.pdf.