



Applied Mathematical Finance I

Lectures 1-2: Introduction and Mathematical Prerequisites

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What is this course about?

- In this course, we will focus on theory and practice of interest rate and credit derivatives.
- To give some flavour, here are some questions that we are going to answer during the course
 - What are common types of interest rate and credit derivatives?
 - How are they valued and hedged?
 - What are popular interest rate models and how are they calibrated in practice?
 - What changed with the global financial crisis of 2007-08?
 - In view of LIBOR Transition, what is the new generation of products and models?



Motivation

- Consider a derivative contract with final payoff depending on a future realization of the Key Rate of the Central Bank of Russia. How it can be hedged?
- In case of equity derivative, a trader would hedge market risk by taking an offsetting position in underlying stock (delta-hedging).
- Interest rate, however, does not represent a price of a traded asset so one cannot directly buy or sell interest rate for hedging purposes.
- Our goal is to develop arbitrage-free pricing theory for fixed-income markets and today we briefly discuss some fundamentals of derivatives pricing.



General Setup

- We consider an economy with continuous and frictionless trading activity over a finite time interval $[0, T]$.
- Uncertainty is modelled via a probability space $(\Omega, \mathcal{F}, \mathbb{P})$.
- Information flow is represented by a filtration $\mathbb{F} = \{\mathcal{F}_t, 0 \leq t \leq T\}$, a family of sub- σ -algebras such that $\mathcal{F}_s \subseteq \mathcal{F}_t$ for $s \leq t$ and $\mathcal{F} = \mathcal{F}_T$. For technical reasons, we require that the filtration satisfies the usual conditions^{*}.
- The state of economy at time t is given by a vector of prices of n dividend-free assets $S_t = (S_t^1, \dots, S_t^n)^\top$.
- Process S is assumed to be adapted to \mathbb{F} meaning that S_t is fully observable at t .

^{*}That is, \mathcal{F}_t is right-continuous for $t \in [0, T]$ and \mathcal{F}_0 contains all subsets of sets of zero probability.



Trading Strategies

- A trading strategy is a progressively measurable process $\phi = (\phi^1, \dots, \phi^n)^\top$. Interpretation: ϕ_t^i denotes the number of units of the i -th asset held in the portfolio at time t .
- The value V_t of the trading strategy ϕ at time t is defined as $V_t = \phi_t^\top \cdot S_t$.
- Trading gains (or losses) over a small time interval $[t, t + dt]$ are $\phi_t^\top \cdot (S_{t+dt} - S_t)$.
- When continuously rebalancing the portfolio over time interval $[t, T]$, the total profit and loss (P&L) become

$$\text{P\&L}(t, T) = \int_t^T \phi_u^\top dS_u.$$

- To proceed, let us briefly recall some relevant mathematical concepts.

Driving Process

- We will consider a special case of the abstract setup above where information is generated by standard k -dimensional Wiener process $W_t = (W_t^1, \dots, W_t^k)^\top$
 - W^i is independent of W^j for $i \neq j$.
 - Independent Gaussian increments $W_t^i - W_s^i \sim \mathcal{N}(0, t - s)$ for $s \leq t$.
 - Each component W^i is a continuous process starting at 0.
- Normally, filtration \mathbb{F} is the one generated by W (possibly augmented)

$$\mathcal{F}_t = \mathcal{F}_t^W = \sigma\{W_s, 0 \leq s \leq t\}.$$



Martingales and Local Martingales

- Let X be a real-valued adapted process with $\mathbb{E}^{\mathbb{P}}|X_t| < \infty$ for all $t \in [0, T]$, where $\mathbb{E}^{\mathbb{P}}$ denotes expectation under measure \mathbb{P} . X is called a \mathbb{P} -martingale on $[0, T]$ if

$$\mathbb{E}_s^{\mathbb{P}}[X_t] = \mathbb{E}^{\mathbb{P}}[X_t \mid \mathcal{F}_s] = X_s \quad \text{a. s.}$$

for all $0 \leq s \leq t \leq T$.

- Interpretation: “best” prediction of a future state X_t is its current value X_s .
- The notion of a martingale can be generalized in the following way. X is said to be a local martingale (under \mathbb{P}) if there exists a sequence of stopping times $\{\tau_n\}_{n=1}^{\infty}$, with $\tau_n \rightarrow T$ a. s. when $n \rightarrow \infty$, such that $X_{t \wedge \tau_n} = X_{\min\{t, \tau_n\}}$ is a martingale.
- Note that local martingales are not necessarily integrable.



Quadratic Variation and Covariation

- For a stochastic process $X : \mathbb{R}_+ \times \Omega \rightarrow \mathbb{R}$, quadratic variation $\langle X \rangle_t$ over $[0, t]$ is defined as

$$\langle X \rangle_t = \lim_{\|P_n\| \rightarrow 0} \sum_{i=1}^n (X_{t_i} - X_{t_{i-1}})^2,$$

where $P_n = \{t_0, t_1, \dots, t_n\}$ is a partition of the interval $[0, t]$ and $\|P_n\| = \max_{1 \leq i \leq n} (t_i - t_{i-1})$ is the mesh. It is only required that the limit exists under convergence in probability.

- Recall that for standard one-dimensional Brownian motion B we have $\langle B \rangle_t = t$.
- Quadratic covariation $\langle X, Y \rangle_t$ of two processes X and Y can be defined via polarization identity

$$\langle X, Y \rangle_t = \frac{1}{2} (\langle X + Y \rangle_t - \langle X \rangle_t - \langle Y \rangle_t).$$



Stochastic Integral

- As we have already seen, P&L of a portfolio over interval $[t, T]$ is given by

$$\text{P\&L} = \int_t^T \phi_u^\top dS_u$$

and hence stochastic integration plays a central role in quantitative finance.

- How do we define stochastic integral $\int_0^t \sigma_s dB_s$, where B is standard one-dimensional Brownian motion?
- First thing to note, is that it cannot be defined pathwise (for fixed $\omega \in \Omega$) as a Riemann–Stieltjes integral because of infinite variation of B on $[0, t]$

$$\sum_{i=1}^n |B_{t_i} - B_{t_{i-1}}| \geq \frac{1}{\max_i |B_{t_i} - B_{t_{i-1}}|} \cdot \sum_{i=1}^n (B_{t_i} - B_{t_{i-1}})^2 \rightarrow \infty \text{ a. s. for } \|P_n\| \rightarrow 0.$$



Stochastic Integral (continued)

- While stochastic integral cannot be defined as the almost sure limit, we can resort to other notions of convergence of random variables.
- Now consider for the moment $\int_0^t B_s dB_s$. Define

$$L_n = \sum_{i=1}^n B_{t_{i-1}} (B_{t_i} - B_{t_{i-1}}), \quad R_n = \sum_{i=1}^n B_{t_i} (B_{t_i} - B_{t_{i-1}}).$$

- Note that

$$L^2 - \lim(R_n - L_n) = L^2 - \lim \sum_{i=1}^n (B_{t_k} - B_{t_{k-1}})^2 = \langle B \rangle_t = t \neq 0.$$

- Therefore, we expect stochastic integral to depend on the choice of points at which the integrand is evaluated. What choice should we make?



Itô Integral

- Recall that P&L of a trading strategy ϕ over $[t, t + dt]$ is $\phi_t^\top \cdot (S_{t+dt} - S_t)$.
- This suggests using left endpoint of each subinterval as the evaluation point of the integrand which leads to the notion of the Itô Integral.
- Itô integral $\int_0^t \sigma_s dB_s$ can be defined for any real-valued progressively measurable process σ such that $\mathbb{P} \left(\int_0^t |\sigma_s|^2 ds < \infty \right) = 1$. Note that in general case construction is rather involved.
- For our needs, it is sufficient to assume that σ is a càdlàg* adapted process with $\mathbb{P} \left(\int_0^t |\sigma_s|^2 ds < \infty \right) = 1$ in which case Itô integral can be seen as

$$\int_0^t \sigma_s dB_s = \lim_{\|P_n\| \rightarrow 0} \sum_{i=1}^n \sigma_{t_{i-1}} (B_{t_i} - B_{t_{i-1}}),$$

where the limit is understood in the sense of convergence in probability.

*That is, almost all sample paths are right-continuous with finite left-hand limits.

Multidimensional Itô Integral

- Let $\sigma : \mathbb{R}_+ \times \Omega \rightarrow \mathbb{R}^{n \times k}$ be an adapted process such that for all $t \in [0, T]$

$$\int_0^t |\sigma_s|^2 ds < \infty,$$

where $|\sigma_s| = \|\sigma_s\|_F = \sqrt{\text{tr}(\sigma_s \cdot \sigma_s^T)}$ is the Frobenius norm. In this case we say that σ is in $\mathcal{L}_{\text{ad}}^2$.

- Now define

$$X_t = \int_0^t \sigma_s dW_s = \int_0^t \begin{pmatrix} \sigma_s^{11} & \dots & \sigma_s^{1k} \\ \vdots & \ddots & \vdots \\ \sigma_s^{n1} & \dots & \sigma_s^{nk} \end{pmatrix} \begin{pmatrix} dW_s^1 \\ \vdots \\ dW_s^k \end{pmatrix}$$

as a vector-valued process with $X_t^i = \sum_{j=1}^k \int_0^t \sigma_s^{ij} dW_s^j$.



Properties of multidimensional Itô Integral

- Define $I_t = \int_0^t \sigma_s dW_s$, where $\sigma : \mathbb{R}_+ \times \Omega \rightarrow \mathbb{R}^{n \times k}$ is in $\mathcal{L}_{\text{ad}}^2$. We have
 - I_t is \mathcal{F}_t -measurable.
 - I_t is a continuous local martingale.
- Define a space $\mathcal{H}_{\text{ad}}^2 \subset \mathcal{L}_{\text{ad}}^2$ as a set of processes σ such that $\mathbb{E}^{\mathbb{P}} \left(\int_0^T |\sigma_s|^2 ds \right) < \infty$. We then have for $\sigma \in \mathcal{H}_{\text{ad}}^2$
 - I_t is a true martingale and, in particular, $\mathbb{E}^{\mathbb{P}} I_t = 0$ for all $t \in [0, T]$.
 - Itô isometry: $\mathbb{E}^{\mathbb{P}} |I_t|^2 = \mathbb{E}^{\mathbb{P}} \left(\int_0^t |\sigma_s|^2 ds \right)$.
 - Auto-covariance: $\mathbb{E}^{\mathbb{P}} (I_t \cdot I_s^T) = \mathbb{E}^{\mathbb{P}} \left(\int_0^{\min\{t,s\}} \sigma_u \cdot \sigma_u^T du \right)$.



Itô Process

- An adapted continuous n -dimensional process X is called an Itô process if it admits a representation

$$X_t = X_0 + \int_0^t \mu_s ds + \int_0^t \sigma_s dW_s, \quad \text{a.s., } \forall t \in [0, T], \quad (1)$$

where stochastic processes $\mu : \mathbb{R}_+ \times \Omega \rightarrow \mathbb{R}^n$ and $\sigma : \mathbb{R}_+ \times \Omega \rightarrow \mathbb{R}^{n \times k}$ are assumed to satisfy suitable integrability conditions.

- It is customary to represent the above formula using the differential notation as

$$dX_t = \mu_t dt + \sigma_t dW_t.$$

- Integration with respect to Itô process is defined via

$$\int_0^t Y_s dX_s = \int_0^t Y_s \mu_s ds + \int_0^t Y_s \sigma_s dW_s,$$

where $Y : \mathbb{R}_+ \times \Omega \rightarrow \mathbb{R}^{m \times n}$.



Itô's Lemma: One-dimensional Case

- Given a function $f : [0, T] \times \mathbb{R} \rightarrow \mathbb{R}$ and a scalar Itô process X

$$dX_t = \mu_t dt + \sigma_t dW_t,$$

the question is whether the process $Y_t = f(t, X_t)$ is an Itô process.

- Suppose that $f \in C^{1,2}$. Then $Y_t = f(t, X_t)$ is an Itô process such that

$$dY_t = \frac{\partial f}{\partial t}(t, X_t) dt + \frac{\partial f}{\partial x}(t, X_t) dX_t + \frac{1}{2} \frac{\partial^2 f}{\partial x^2}(t, X_t) d\langle X \rangle_t,$$

where quadratic variation of X is given by

$$\langle X \rangle_t = \int_0^t \sigma_s^2 ds.$$

- We usually simply write the above relation as $dX_t^2 = \sigma_t^2 dt$.



Itô's Lemma: Multidimensional Case

- Now consider $f : [0, T] \times \mathbb{R}^n \rightarrow \mathbb{R}$, $f \in C^{1,2}$, and assume that X is given by formula (1). Then $Y_t = f(t, X_t)$ is an Itô process such that

$$dY_t = \frac{\partial f}{\partial t}(t, X_t) dt + \sum_{i=1}^n \frac{\partial f}{\partial x_i}(t, X_t) dX_t^i + \frac{1}{2} \sum_{i,j=1}^n \frac{\partial^2 f}{\partial x_i \partial x_j}(t, X_t) d\langle X^i, X^j \rangle_t,$$

where

$$\langle X^i, X^j \rangle_t = \int_0^t \sum_{l=1}^k \sigma_s^{il} \sigma_s^{jl} ds.$$

- In particular, taking $f(x_1, x_2) = x_1 x_2$, we get Itô product rule for scalar processes X and Y

$$d(X_t Y_t) = X_t dY_t + Y_t dX_t + d\langle X, Y \rangle_t.$$



Stochastic Differential Equations

- Usually, we restrict our attention to the case where both coefficients μ and σ in (1) are deterministic measurable functions of time and state.
- In other words, we consider stochastic differential equation (SDE) of the form

$$dX_t = \mu(t, X_t) dt + \sigma(t, X_t) dW_t, \quad X_0 = x \in \mathbb{R}^n.$$

- A strong solution to the SDE is an Itô process X adapted to \mathcal{F}_t^W and satisfying

$$X_t = x + \int_0^t \mu(s, X_s) ds + \int_0^t \sigma(s, X_s) dW_s.$$

- As is well known, if $\mu(t, x)$ and $\sigma(t, x)$ satisfy the Lipschitz and linear growth condition in x , then there exists a unique solution to the SDE above. Moreover, it is a Markov process.



Self-financing Strategies

- Now that the basics are covered, let us get back to our setup.
- A trading strategy ϕ is said to be self-financing if, for any $t \in [0, T]$, its value $V_t = \phi_t^\top \cdot S_t$ satisfy

$$V_t - V_0 = \int_0^t \phi_u^\top dS_u,$$

or, in differential notation,

$$dV_t = \phi_t^\top dS_t.$$

- Interpretation: changes in portfolio value are only due to trading gains or losses, with no funds being added or withdrawn.
- Note that Itô product rule implies

$$S_t^\top d\phi_t + d\phi_t^\top dS_t = 0.$$



Arbitrage

- By arbitrage we generally mean a possibility to create “something for nothing”.
- Let us denote by Φ the class of all self-financing strategies. Formally, an arbitrage opportunity is a self-financing strategy $\phi \in \Phi$ for which $V_0 = 0$ and, for some $t \in [0, T]$,

$$\mathbb{P}(V_t \geq 0) = 1 \quad \text{and} \quad \mathbb{P}(V_t > 0) > 0$$

or, equivalently,

$$\mathbb{P}(V_t \geq 0) = 1 \quad \text{and} \quad \mathbb{E}^{\mathbb{P}} V_t > 0.$$

- In economic equilibrium, arbitrage should not exist and requiring the absence of arbitrage strategies is a starting point of asset pricing.



Market Model

- We assume the dynamics of the price vector S are given by an Itô process

$$dS_t = \mu_t dt + \sigma_t dW_t, \quad S_0 = s.$$

- By market model we mean a pair (S, Ψ) , where $\Psi \subseteq \Phi$ is a set of all “allowable” self-financing trading strategies.
- It turns out that arbitrage opportunities are not excluded a priori from the class of self-financing strategies and hence (S, Φ) is generally not free of arbitrage even for “good” choice of price dynamics.
- This means that we have to impose some additional restrictions on the set of allowed trading strategies to get a meaningful model.
- We now turn to the question of characterizing the conditions under which market model is arbitrage-free.



Equivalent Measures and Radon-Nikodym Theorem

- First, we recall that two probability measures \mathbb{P} and $\hat{\mathbb{P}}$ (on the same measurable space) are said to be equivalent ($\mathbb{P} \sim \hat{\mathbb{P}}$) if they have the same null-sets

$$\mathbb{P}(B) = 0 \iff \hat{\mathbb{P}}(B) = 0, \quad \forall B \in \mathcal{F}.$$

- Radon-Nikodym Theorem

Suppose that $\hat{\mathbb{P}} \sim \mathbb{P}$. There exists a unique (a.s.) random variable ξ , which is called a Radon-Nikodym derivative and is usually denoted as $\frac{d\hat{\mathbb{P}}}{d\mathbb{P}}$, such that

$$\hat{\mathbb{P}}(B) = \int_B \xi \, d\mathbb{P}, \quad \forall B \in \mathcal{F}.$$

Of course, we must have $\frac{d\hat{\mathbb{P}}}{d\mathbb{P}} > 0$ a.s. and $\mathbb{E}^{\mathbb{P}} \left[\frac{d\hat{\mathbb{P}}}{d\mathbb{P}} \right] = 1$.

Properties of Radon-Nikodym Derivative

- Assume that $\hat{\mathbb{P}} \sim \mathbb{P}$ and let \mathcal{X} be a $\hat{\mathbb{P}}$ -integrable random variable. Then

$$\mathbb{E}^{\hat{\mathbb{P}}} \mathcal{X} = \mathbb{E}^{\mathbb{P}} \left[\frac{d\hat{\mathbb{P}}}{d\mathbb{P}} \mathcal{X} \right].$$

- Also, we have $\frac{d\mathbb{P}}{d\hat{\mathbb{P}}} \frac{d\hat{\mathbb{P}}}{d\mathbb{P}} = 1$ a.s. To see this, we note that for any $B \in \mathcal{F}$

$$\mathbb{P}(B) = \int_B 1 \, d\mathbb{P} = \int_B \frac{d\mathbb{P}}{d\hat{\mathbb{P}}} \, d\hat{\mathbb{P}} = \int_B \frac{d\mathbb{P}}{d\hat{\mathbb{P}}} \frac{d\hat{\mathbb{P}}}{d\mathbb{P}} \, d\mathbb{P}.$$



Numéraires and Equivalent Martingale Measures

- It is usually convenient to value assets relatively to some benchmark asset called a numéraire.
- A numéraire N is any self-financing strategy such that its value N_t is strictly positive (a.s.) for all $t \in [0, T]$.
- A measure $\mathbb{Q}^N \sim \mathbb{P}$ is called an equivalent martingale measure induced by N if the normalized asset process $\frac{S}{N}$ is a \mathbb{Q}^N -martingale. Note that such measure is not necessarily unique.
- If \mathbb{Q}^N exists, (N, \mathbb{Q}^N) is referred to as a numéraire pair.

Numéraires and Self-financing Strategies

- Consider a numéraire pair (N, \mathbb{Q}^N) . A normalized price $\frac{V}{N}$ of an arbitrary self-financing portfolio ϕ will generally only be a local martingale under \mathbb{Q}^N .
- Indeed, applying Itô product rule combined with the definition of a self-financing portfolio, we have

$$\begin{aligned} d\left(\frac{V_t}{N_t}\right) &= \frac{1}{N_t} dV_t + V_t d\left(\frac{1}{N_t}\right) + dV_t d\left(\frac{1}{N_t}\right) \\ &= \frac{1}{N_t} \phi_t^\top dS_t + \phi_t^\top S_t d\left(\frac{1}{N_t}\right) + \phi_t^\top dS_t d\left(\frac{1}{N_t}\right) \\ &= \phi_t^\top d\left(\frac{S_t}{N_t}\right). \end{aligned}$$

- The fact that $\frac{V}{N}$ is a local martingale now follows from Martingale Representation Theorem and properties of Itô integral.



Martingale Representation Theorem

- It follows from properties of Itô integral that all driftless Itô processes of the type

$$dX_t = \sigma_t dW_t$$

are local martingales.

- A converse result is also true: if X_t is a local martingale adapted to filtration generated by W , then there exists a process σ such that the above formula holds.
- Corollary: any integrable random variable \mathcal{X} measurable with respect to \mathcal{F}_T^W can be represented as

$$\mathcal{X} = \mathbb{E}^{\mathbb{P}} \mathcal{X} + \int_0^T \gamma_t dW_t$$

for some process γ .



First Fundamental Theorem of Asset Pricing

- Let N be a numéraire and \mathbb{Q}^N be a corresponding equivalent martingale measure.
- We are now ready to define “allowable” strategies. Self-financing strategy ϕ is called permissible if its normalized price process $\frac{V}{N}$ is a true \mathbb{Q}^N -martingale. We denote by $\Psi(N)$ the set of all permissible strategies.
- Sufficient Condition for No-Arbitrage
If there exists a martingale measure \mathbb{Q}^N corresponding to some numéraire N , then market model $(S, \Psi(N))$ is free of arbitrage.



First Fundamental Theorem of Asset Pricing: Proof

- Consider any permissible strategy $\phi \in \Psi(N)$ such that $V_0 = 0$ and $\mathbb{P}(V_t \geq 0) = 1$ for some time t .
- Given that $\mathbb{Q}^N \sim \mathbb{P}$, we first note that $\mathbb{Q}^N \left(\frac{V_t}{N_t} \geq 0 \right) = 1$. Since $\frac{V}{N}$ is a \mathbb{Q}^N -martingale, we also have

$$\mathbb{E}^{\mathbb{Q}^N} \left[\frac{V_t}{N_t} \right] = \mathbb{E}^{\mathbb{Q}^N} \left[\frac{V_0}{N_0} \right] = 0$$

and therefore

$$\mathbb{Q}^N \left(\frac{V_t}{N_t} = 0 \right) = 1.$$

- The above result implies that $\mathbb{P}(V_t = 0) = 1$ which means that ϕ is not an arbitrage opportunity and hence $\Psi(N)$ is free of arbitrage.



Derivative Securities

- A derivative security (contingent claim) with maturity T is an \mathcal{F}_T -measurable random variable \mathcal{X}_T . We will assume that \mathcal{X}_T has a finite variance.
- Interpretation: stochastic payoff at T and no payments before T .
- We say that derivative security \mathcal{X}_T is attainable if there exists a permissible trading strategy ϕ such that

$$V_T = \phi_T^\top \cdot S_T = \mathcal{X}_T \quad \text{a.s.}$$

- The trading strategy ϕ in the above formula is said to replicate the derivative security.



Derivatives Pricing

- Consider a numéraire N inducing a martingale measure \mathbb{Q}^N . Let π_t denote the price of an attainable contingent claim \mathcal{X}_T at time t .
- No-arbitrage principle implies that π_t must be equal to the value of the replicating portfolio V_t . Since the replicating portfolio is not yet known explicitly (we only know it exists), how do we actually compute π_t ?
- By the First Fundamental Theorem of Asset Pricing

$$\frac{\pi_t}{N_t} = \frac{V_t}{N_t} = \mathbb{E}_t^{\mathbb{Q}^N} \left[\frac{V_T}{N_T} \right] = \mathbb{E}_t^{\mathbb{Q}^N} \left[\frac{\mathcal{X}_T}{N_T} \right]$$

and therefore

$$\pi_t = N_t \cdot \mathbb{E}_t^{\mathbb{Q}^N} \left[\frac{\mathcal{X}_T}{N_T} \right].$$



Complete Markets

- If all contingent claims \mathcal{X}_T can be replicated, the market model is said to be complete.
- Interpretation: in complete markets derivative securities are redundant in a sense that they can be created synthetically by trading the underlying assets S^1, \dots, S^n .
- Second Fundamental Theorem of Asset Pricing
In the absence of arbitrage, a market model is complete if and only if there exists a numéraire N inducing the unique martingale measure \mathbb{Q}^N .



Abstract Bayes' Formula

- Consider two equivalent probability measures $\hat{\mathbb{P}} \sim \mathbb{P}$ on (Ω, \mathcal{F}_T) and let \mathcal{X} be an \mathcal{F}_T -measurable random variable integrable with respect to $\hat{\mathbb{P}}$. Then, for $t \in [0, T]$,

$$\mathbb{E}_t^{\hat{\mathbb{P}}} \mathcal{X} = \frac{\mathbb{E}_t^{\mathbb{P}} \left[\frac{d\hat{\mathbb{P}}}{d\mathbb{P}} \mathcal{X} \right]}{\mathbb{E}_t^{\mathbb{P}} \left[\frac{d\hat{\mathbb{P}}}{d\mathbb{P}} \right]}. \quad (2)$$

- By definition, $\mathbb{E}_t^{\hat{\mathbb{P}}} \mathcal{X}$ is a random variable satisfying the following conditions
 - $\mathbb{E}_t^{\hat{\mathbb{P}}} \mathcal{X}$ is \mathcal{F}_t -measurable
 - For any event $A \in \mathcal{F}_t$

$$\int_A \mathbb{E}_t^{\hat{\mathbb{P}}} \mathcal{X} d\hat{\mathbb{P}} = \int_A \mathcal{X} d\hat{\mathbb{P}}.$$

- Also, recall that for unconditional expectations we have

$$\mathbb{E}^{\hat{\mathbb{P}}} \mathcal{X} = \mathbb{E}^{\mathbb{P}} \left[\frac{d\hat{\mathbb{P}}}{d\mathbb{P}} \mathcal{X} \right].$$

Abstract Bayes' Formula: Proof

- We rewrite formula (2) as

$$\mathbb{E}_t^{\mathbb{P}} \left[\frac{d\hat{\mathbb{P}}}{d\mathbb{P}} \mathcal{X} \right] = \mathbb{E}_t^{\hat{\mathbb{P}}} \mathcal{X} \cdot \mathbb{E}_t^{\mathbb{P}} \left[\frac{d\hat{\mathbb{P}}}{d\mathbb{P}} \right].$$

- The right-hand side of the last formula is \mathcal{F}_t -measurable, so it is enough to show

$$\int_A \frac{d\hat{\mathbb{P}}}{d\mathbb{P}} \cdot \mathcal{X} d\mathbb{P} = \int_A \mathbb{E}_t^{\hat{\mathbb{P}}} \mathcal{X} \cdot \mathbb{E}_t^{\mathbb{P}} \left[\frac{d\hat{\mathbb{P}}}{d\mathbb{P}} \right] d\mathbb{P}, \quad \forall A \in \mathcal{F}_t.$$

- We have

$$\begin{aligned} \int_A \frac{d\hat{\mathbb{P}}}{d\mathbb{P}} \cdot \mathcal{X} d\mathbb{P} &= \int_A \mathcal{X} d\hat{\mathbb{P}} = \int_A \mathbb{E}_t^{\hat{\mathbb{P}}} \mathcal{X} d\hat{\mathbb{P}} = \int_A \mathbb{E}_t^{\hat{\mathbb{P}}} \mathcal{X} \cdot \frac{d\hat{\mathbb{P}}}{d\mathbb{P}} d\mathbb{P} \\ &= \int_A \mathbb{E}_t^{\mathbb{P}} \left[\mathbb{E}_t^{\hat{\mathbb{P}}} \mathcal{X} \cdot \frac{d\hat{\mathbb{P}}}{d\mathbb{P}} \right] d\mathbb{P} = \int_A \mathbb{E}_t^{\hat{\mathbb{P}}} \mathcal{X} \cdot \mathbb{E}_t^{\mathbb{P}} \left[\frac{d\hat{\mathbb{P}}}{d\mathbb{P}} \right] d\mathbb{P}. \end{aligned}$$



Radon-Nikodym Density Process

- For two equivalent probability measures $\hat{\mathbb{P}} \sim \mathbb{P}$ on (Ω, \mathcal{F}_T) , a density process ξ associated with $\hat{\mathbb{P}}$ is defined as

$$\xi_t = \mathbb{E}_t^{\mathbb{P}} \left[\frac{d\hat{\mathbb{P}}}{d\mathbb{P}} \right], \quad t \in [0, T].$$

- Note that ξ is a \mathbb{P} -martingale

$$\mathbb{E}_s^{\mathbb{P}} \xi_t = \mathbb{E}_s^{\mathbb{P}} \left[\mathbb{E}_t^{\mathbb{P}} \left[\frac{d\hat{\mathbb{P}}}{d\mathbb{P}} \right] \right] = \mathbb{E}_s^{\mathbb{P}} \left[\frac{d\hat{\mathbb{P}}}{d\mathbb{P}} \right] = \xi_s, \quad s \leq t.$$

- If ζ is the density process for \mathbb{P} , i.e., $\zeta_t = \mathbb{E}_t^{\hat{\mathbb{P}}} \left[\frac{d\mathbb{P}}{d\hat{\mathbb{P}}} \right]$, then $\zeta = \frac{1}{\xi}$. Indeed

$$\zeta_t = \mathbb{E}_t^{\hat{\mathbb{P}}} \left[\frac{d\mathbb{P}}{d\hat{\mathbb{P}}} \right] = \frac{\mathbb{E}_t^{\mathbb{P}} \left[\frac{d\mathbb{P}}{d\hat{\mathbb{P}}} \frac{d\hat{\mathbb{P}}}{d\mathbb{P}} \right]}{\mathbb{E}_t^{\mathbb{P}} \left[\frac{d\hat{\mathbb{P}}}{d\mathbb{P}} \right]} = \frac{1}{\mathbb{E}_t^{\mathbb{P}} \left[\frac{d\hat{\mathbb{P}}}{d\mathbb{P}} \right]} = \frac{1}{\xi_t}.$$



Radon-Nikodym Density Process (continued)

- Consider a measurable space (Ω, \mathcal{F}_t) for $t < T$. Define $\mathbb{P}_t = \mathbb{P}|_{\mathcal{F}_t}$ and $\hat{\mathbb{P}}_t = \hat{\mathbb{P}}|_{\mathcal{F}_t}$.
- There exists \mathcal{F}_t -measurable random variable $\frac{d\hat{\mathbb{P}}_t}{d\mathbb{P}_t}$ such that

$$\hat{\mathbb{P}}_t(B) = \int_B \frac{d\hat{\mathbb{P}}_t}{d\mathbb{P}_t} d\mathbb{P}_t, \quad \forall B \in \mathcal{F}_t.$$

- We note that $\frac{d\hat{\mathbb{P}}_t}{d\mathbb{P}_t}$ is given by

$$\frac{d\hat{\mathbb{P}}_t}{d\mathbb{P}_t} = \mathbb{E}_t^{\mathbb{P}} \left[\frac{d\hat{\mathbb{P}}}{d\mathbb{P}} \right].$$

Indeed, for any $B \in \mathcal{F}_t$

$$\int_B \frac{d\hat{\mathbb{P}}_t}{d\mathbb{P}_t} d\mathbb{P}_t = \hat{\mathbb{P}}_t(B) = \int_B 1 d\hat{\mathbb{P}}_t = \int_B \frac{d\hat{\mathbb{P}}}{d\mathbb{P}} d\mathbb{P}_t.$$



Change of Numéraire

- Consider a numéraire pair (N, \mathbb{Q}^N) . Recall that the arbitrage-free price at time t of an attainable contingent claim \mathcal{X}_T is given by

$$\pi_t = N_t \cdot \mathbb{E}_t^{\mathbb{Q}^N} \left[\frac{\mathcal{X}_T}{N_T} \right]. \quad (3)$$

- Now let us consider another permissible numéraire M . Note that $\frac{M}{N}$ is a \mathbb{Q}^N -martingale and, in particular, $\mathbb{E}^{\mathbb{Q}^N} \left[\frac{M_T}{N_T} \right] = \frac{M_0}{N_0}$.
- Define a new equivalent probability measure \mathbb{Q}^M on (Ω, \mathcal{F}_T) via specifying a Radon-Nikodym derivative ξ_T

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

- Note that \mathbb{Q}^M is indeed a probability measure since $\xi_T > 0$ a.s. and $\mathbb{E}^{\mathbb{Q}^N} [\xi_T] = 1$.

Change of Numéraire (continued)

- By applying abstract Bayes' formula to (3), we get

$$\pi_t = N_t \cdot \mathbb{E}_t^{\mathbb{Q}^N} \left[\frac{\mathcal{X}_T}{N_T} \right] = N_t \cdot \mathbb{E}_t^{\mathbb{Q}^N} [\xi_T] \cdot \mathbb{E}_t^{\mathbb{Q}^M} \left[\frac{\mathcal{X}_T}{\xi_T \cdot N_T} \right].$$

- Again, using the fact that $\frac{M}{N}$ is a martingale under \mathbb{Q}^N , we have

$$\begin{aligned} \pi_t &= N_t \cdot \frac{M_t}{N_t} \cdot \frac{N_0}{M_0} \cdot \mathbb{E}_t^{\mathbb{Q}^M} \left[\frac{\mathcal{X}_T \cdot M_0 \cdot N_T}{M_T \cdot N_0 \cdot N_T} \right] \\ &= M_t \cdot \mathbb{E}_t^{\mathbb{Q}^M} \left[\frac{\mathcal{X}_T}{M_T} \right]. \end{aligned}$$

- This is a very powerful technique since, for a particular payoff \mathcal{X}_T , we can try to find a suitable numéraire M so that $\mathbb{E}_t^{\mathbb{Q}^M} \left[\frac{\mathcal{X}_T}{M_T} \right]$ is easy to evaluate.



Doléans-Dade Exponential

- Consider a k -dimensional process θ adapted to filtration generated by Wiener process $W = W^{\mathbb{P}}$. Define a scalar random process Z_t

$$dZ_t = Z_t \theta_t^{\top} dW^{\mathbb{P}}, \quad Z_0 = 1. \quad (4)$$

- By an application of Itô's lemma, it can be easily verified that

$$\begin{aligned} Z_t &= e^{-\frac{1}{2} \int_0^t \theta_s^{\top} \theta_s ds + \int_0^t \theta_s^{\top} dW_s^{\mathbb{P}}} \\ &= \mathcal{E} \left(\int_0^t \theta_s^{\top} dW_s^{\mathbb{P}} \right), \end{aligned}$$

where $\mathcal{E}(\cdot)$ is the Doléans-Dade exponential.

- Recall that for a continuous semimartingale X , Doléans-Dade exponential is defined as

$$\mathcal{E}(X)_t = e^{X_t - X_0 - \frac{1}{2} \langle X \rangle_t}.$$



Girsanov Theorem

- The most important tool when dealing with equivalent measure transformation is Girsanov theorem.
- Suppose that process Z in (4) is a martingale. Then process $W^{\hat{\mathbb{P}}}$ given by

$$W_t^{\hat{\mathbb{P}}} = W_t^{\mathbb{P}} - \int_0^t \theta_s ds, \quad t \in [0, T],$$

is a Wiener process under probability measure $\hat{\mathbb{P}}$ defined via its Radon-Nikodym derivative $\frac{d\hat{\mathbb{P}}}{d\mathbb{P}} = Z_T$.

- Sufficient condition for process Z to be a martingale is the Novikov's condition

$$\mathbb{E}^{\mathbb{P}} \left[e^{\frac{1}{2} \int_0^T \theta_t^T \theta_t dt} \right] < \infty.$$



Girsanov Theorem and Effect on Asset Dynamics

- Recall that asset price process dynamics under \mathbb{P} are given by

$$dS_t = \mu_t dt + \sigma_t dW_t^{\mathbb{P}}.$$

- We now consider a measure change from \mathbb{P} to $\hat{\mathbb{P}} \sim \mathbb{P}$ given by the exponential martingale kernel Z_T .
- By Girsanov theorem, dynamics of $W^{\mathbb{P}}$ under $\hat{\mathbb{P}}$ become

$$dW_t^{\mathbb{P}} = dW_t^{\hat{\mathbb{P}}} + \theta_t dt.$$

which implies the following $\hat{\mathbb{P}}$ -dynamics of S

$$dS_t = (\mu_t + \sigma_t \cdot \theta_t) dt + \sigma_t dW_t^{\hat{\mathbb{P}}}.$$



Markovian Setup

- In a complete market, any contingent claim can be replicated by a self-financing trading strategy, but it still remains to determine that strategy explicitly.
- Let us now consider a Markovian setup where price process dynamics under \mathbb{P} are given by

$$dS_t = \mu(t, S_t)dt + \sigma(t, S_t)dW_t^{\mathbb{P}}.$$

- Consider an attainable derivative security with payoff of the form $\mathcal{X}_T = F(S_T)$. The Markovian form of the asset dynamics suggests that the derivative price π_t is a function of t and S_t only, $\pi_t = V(t, S_t)$ for some deterministic function $V(t, s)$.



Derivatives Pricing and PDEs

- Assuming that function V above is smooth enough, we apply Itô's lemma to get

$$\begin{aligned} dV(t, S_t) = & \frac{\partial V(t, S_t)}{\partial t} dt + \sum_{i=1}^n \frac{\partial V(t, S_t)}{\partial s_i} \mu^i(t, S_t) dt + \frac{1}{2} \sum_{i,j=1}^n \frac{\partial^2 V(t, S_t)}{\partial s_i \partial s_j} \Sigma_{i,j} dt \\ & + \sum_{i=1}^n \frac{\partial V(t, S_t)}{\partial s_i} \sigma^i(t, S_t) dW_t^{\mathbb{P}}, \end{aligned}$$

where σ^i is the i -th row of the $n \times k$ matrix σ and $\Sigma_{i,j}$ is the (i,j) -th element of the matrix $\sigma \sigma^{\top}$.

- Since contingent claim is attainable, there exists a permissible self-financing replicating portfolio ϕ and hence we must have

$$dV(t, S_t) = \phi_t^{\top} dS_t = \sum_{i=1}^n \phi_t^i \mu^i(t, S_t) dt + \sum_{i=1}^n \phi_t^i \sigma^i(t, S_t) dW_t^{\mathbb{P}}.$$



Derivatives Pricing and PDEs (continued)

- Comparing terms in the two equations above, we see that both equations will hold given that

$$\phi_t^i = \frac{\partial V(t, S_t)}{\partial s_i}, \quad i = 1, \dots, n,$$

and

$$\frac{\partial V(t, s)}{\partial t} + \frac{1}{2} \sum_{i,j=1}^n \frac{\partial^2 V(t, s)}{\partial s_i \partial s_j} \Sigma_{i,j} = 0, \quad V(T, s) = F(s). \quad (5)$$

- What it tells us is that
 - Value function $V(t, s)$ solves the second-order parabolic Cauchy problem (5).
 - To replicate derivative contract paying $F(S_T)$ at time T , one must hold $\frac{\partial V(t, S_t)}{\partial s_i}$ units of asset S^i in portfolio at t .



Black-Scholes-Merton Model

- In the Black-Scholes-Merton model, the economy consists of two assets: the bank (money market) account B and the stock S .
- Dynamics of the bank account are given by

$$dB_t = rB_t dt, \quad B_0 = 1, r \geq 0.$$

Interpretation: cash amount on the account grows at a continuously compounded risk-free interest rate r .

- The stock price dynamics are given by a Geometric Brownian motion

$$dS_t = S_t \left[\mu dt + \sigma dW_t^{\mathbb{P}} \right],$$

where $W^{\mathbb{P}}$ is a one-dimensional Wiener process and μ and $\sigma > 0$ are constants.
Interpretation: instantaneous relative returns follow Brownian motion with drift.



Black-Scholes-Merton Model: Probabilistic Approach

- Consider a trading strategy $\phi = (\phi_t^1, \phi_t^2)_{t \in [0, T]}^\top \equiv (1, 0)^\top$ which corresponds to depositing one unit of cash at a bank account and holding it there until time T .
- Obviously, ϕ is self-financing and its value V_t is given by $V_t = 1 \cdot B_t + 0 \cdot S_t = B_t$. Therefore, B itself is a valid numéraire.
- By applying Itô's lemma to normalized stock price $\frac{S}{B}$, we get

$$d\left(\frac{S_t}{B_t}\right) = \frac{S_t}{B_t} \left[(\mu - r)dt + \sigma dW^\mathbb{P} \right].$$

- In view of Girsanov theorem, it is clear that, by choosing $\theta_t = \frac{r - \mu}{\sigma}$, $t \in [0, T]$, we can define an equivalent martingale measure \mathbb{Q} with Radon-Nikodym derivative being

$$\frac{d\mathbb{Q}}{d\mathbb{P}} = e^{-\frac{1}{2} \frac{(r - \mu)^2}{\sigma^2} T + \frac{r - \mu}{\sigma} W_T^\mathbb{P}}. \quad (6)$$



Black-Scholes-Merton Model: Probabilistic Approach

- The probability measure \mathbb{Q} given by (6) is called the risk-neutral measure.
- Under \mathbb{Q} , $W_t^{\mathbb{Q}} = W_t^{\mathbb{P}} - \frac{r-\mu}{\sigma} t$ is a Wiener process and

$$d\left(\frac{S_t}{B_t}\right) = \frac{S_t}{B_t} \sigma dW_t^{\mathbb{Q}},$$
$$dS_t = S_t \left[r dt + \sigma dW_t^{\mathbb{Q}} \right].$$

- By applying Itô's lemma, it can be easily verified that for stock price S_T we have

$$S_T = S_t e^{r(T-t)} e^{-\frac{1}{2}\sigma^2(T-t) + \sigma(W_T^{\mathbb{Q}} - W_t^{\mathbb{Q}})}, \quad t \in [0, T].$$

- Note that the risk-neutral measure \mathbb{Q} is unique and hence the market is arbitrage-free and complete.



Call Option Pricing in BSM Setting

- Call option pays out an amount of $\max(S_T - K, 0) = (S_T - K)_+$ at T .
- By arbitrage-free pricing formula, call option price π_t at time $t \leq T$ is given by

$$\begin{aligned}\pi_t &= B_t \cdot \mathbb{E}_t^{\mathbb{Q}} \left[\frac{(S_T - K)_+}{B_T} \right] = e^{-r(T-t)} \cdot \mathbb{E}_t^{\mathbb{Q}} [(S_T - K)_+] \\ &= e^{-r(T-t)} \cdot \int_{-\infty}^{\infty} \left(S_t e^{r(T-t)} e^{-\frac{1}{2}\sigma^2(T-t) + \sigma z \sqrt{T-t}} - K \right)_+ \varphi(z) dz,\end{aligned}$$

where $\varphi(z) = \Phi'(z)$, $\Phi(z) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^z e^{-\frac{y^2}{2}} dy$, is the standard Gaussian density.

- In BSM setting, the integral above can be computed explicitly and we get the celebrated Black-Scholes-Merton formula

$$\pi_t = S_t \Phi(d_+) - K e^{-r(T-t)} \Phi(d_-),$$

where $d_+ = \frac{\ln \frac{S_t}{K} + (r + \sigma^2)(T-t)}{\sigma \sqrt{T-t}}$, $d_- = d_+ - \sigma \sqrt{T-t}$.



Call Option Pricing via PDEs

- Our derivation of the call option price was so far entirely probabilistic. Let us now show how one can apply the PDE technique to price a call option.
- Writing $\pi_t = V(t, B_t, S_t)$ for some deterministic function $V(t, b, s)$, we see that pricing PDE (5) becomes

$$\frac{\partial V(t, b, s)}{\partial t} + \frac{1}{2}s^2\sigma^2 \frac{\partial^2 V(t, b, s)}{\partial s^2} = 0,$$

subject to the boundary condition $V(T, b, s) = (s - K)_+$.

- Since $B_t = e^{rt}$ is a deterministic function of t , we can actually introduce function $P(t, s) = V(t, e^{rt}, s)$. The PDE above, when written in terms of P , becomes the famous Black-Scholes-Merton PDE

$$\frac{\partial P(t, s)}{\partial t} + rS \frac{\partial P(t, s)}{\partial s} + \frac{1}{2}s^2\sigma^2 \frac{\partial^2 P(t, s)}{\partial s^2} = rP(t, s).$$



Example

- We would like to price the payoff $(S_T^3 - S_T^2)_+$ in a simple BSM setting with $r = 0$ which implies that $B \equiv 1$ and $dS_t = \sigma S_t dW_t^{\mathbb{Q}}$ under the risk-neutral measure \mathbb{Q} .
- The initial price of the contract is then given by

$$\pi_0 = \mathbb{E}^{\mathbb{Q}} \left[(S_T^3 - S_T^2)_+ \right] = \mathbb{E}^{\mathbb{Q}} \left[S_T^2 (S_T - 1)_+ \right]$$

and it may look like choosing S^2 as a numéraire allows us to simplify the problem to a pricing of a standard call option under martingale measure \mathbb{Q}^{S^2}

$$\pi_0 = S_0^2 \cdot \mathbb{E}^{\mathbb{Q}^{S^2}} [(S_T - 1)_+].$$

- If \mathbb{Q}^{S^2} exists, the normalized price process $\left(\frac{B}{S}, \frac{S}{S^2} \right) = \left(\frac{1}{S^2}, \frac{1}{S} \right)$ must be a \mathbb{Q}^{S^2} -martingale which is impossible. Something went wrong...

