

The Probabilistic Solution

- 我们之前学了一些 Valuation 的方法

- Binomial $V_{j,i} = e^{-r\delta t} (q V_{j+1,i+1} + (1-q) V_{j+1,i})$

- European $V_0 = e^{-rT} (S_0 e^{(r-\delta)T} N(d_1) - K N(d_2))$

- PDE $\frac{\partial V}{\partial t} + \underbrace{(r-\delta)}_{\downarrow} S \frac{\partial V}{\partial S} + \frac{1}{2} \sigma^2 S^2 \frac{\partial^2 V}{\partial S^2} - rV = 0$

Risk-neutral expected return

- 方法已经有了，但我们如何推导出 R-N prob. 呢？
- 我们又如何利用 R-N prob. 来定 options 的价呢？
 - * option values \Rightarrow expectations / solutions of PDEs

下面我们将介绍两大知识三：Martingale + Girsanov's Theorem

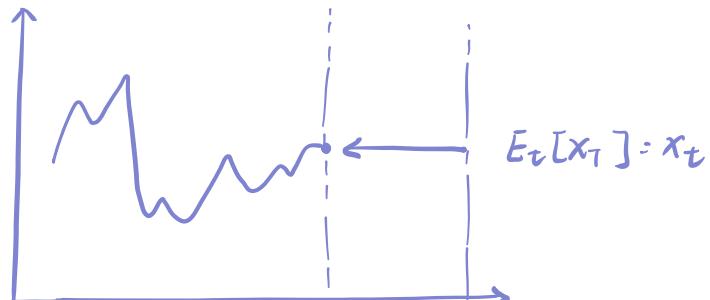
"我们如何从 real world 进入 risk-neutral world 中呢？"

- Martingale

什么是 martingale 呢？

"No drift" 的 stochastic process

★ $E_t[X_T] = x_t (T > t)$



↳ take expectation at time t

注意! In finance very few things are martingales

any financial stochastic variables 基本都不是

(需要期望是0, 像债券, 股票, 期权基本都有正收益)

—要不然我们还买它们干嘛呢?

当 $r=0$ 时 option value $V_{j,i} = \frac{e^{-r\Delta t}}{\downarrow} (qV_{j+1,i+1} + (1-q)V_{j+1,i}) \Downarrow$

$$V_{j,i} = 1 \times E_r [V_{j+1}]$$

Quiz : 下列五项哪项是 martingale

- Stock price, S .
- Option price, V .
- Stock price, S , under risk-neutral measure Q when $R=0$ ✓ → Risk-neutral world 一切资产的收益都是 μ
- Bond price, B , under risk-neutral measure Q . ✗
- The total received, X , in a gambling game where you toss a coin a finite number of times, receiving \$1 for each head and paying \$1 for each tail? → Martingale 名字就是 Martingale ✓

注意! 在 Lecture 2 中, 我们提到且

In the risk-neutral world the mean (μ) and variance (ν) of a normally distributed $r_{\Delta t}$ are:

$$\mu = \left(r - \frac{1}{2}\sigma^2 \right) \Delta t \text{ 由上页得到}$$
$$\nu = \sigma^2 \Delta t \text{ 假设我们知道}$$

那这是怎么回事呢? 注意 $r_{\Delta t}$ 是 $\ln \frac{S_{\Delta t}}{S_0}$ 其目的想变成 $E^{RN}[S_{\Delta t}] = S_0 \cdot e^{\mu \Delta t}$
这的本质是不变的

由上面的 quiz 可以看出, 除非我们刻意设置 $R=0$, 否则一般的金融产品都不是 martingale. — 所以我们学它干嘛?

— 即使 $R \neq 0$, martingale 也是存在的!

— 在 Binomial model 里面有 martingale → Martingale 是存在的

- Martingale 是什么, 在哪里存在, 为什么使用它

Binomial Model, with prob. q

$$B_1 = B_0(1+R) \quad (= B_0 e^r)$$

$$E[S_1] = S_0(1+R) \quad (= S_0 e^r)$$

↓

$$E\left[\frac{S_1}{B_1}\right] = \frac{S_0(1+R)}{B_0(1+R)} = \frac{S_0}{B_0} \Rightarrow \frac{S_t}{B_t} \text{ is a martingale using prob. } q$$

在 Risk-neutral world 中很多比率都是 martingale

* 不是因为是 Martingale 的
那我们为啥要用它呢？因为它有很好的性质 → replicating property

If we have a probability measure α (有一个这样的世界，不同 state 分配

-T prob.)

在这 T α 中，有两 T martingales : N, S



$$N_t = N_0 + \sum_{k=1}^t \Delta S_k \times \phi_k$$

我们总能找到一个系数 ϕ_k ，让

$$(\Delta N_k = \underbrace{\phi_k}_{\text{known parameter}} \cdot \Delta S_k)$$

-T martingale 用另一-T martingale 来表示

In Binomial

$$\phi_k \Leftrightarrow \Delta \quad * 忘了 \Delta 是什么了吗？快翻 PPT 看一看吧！$$

(*) * 这里我也不很确定，可能是指 Replicating portfolio 中 S 的系数

即用 $\alpha S + \beta$ 来换 option value. → 但这和 ϕ_k 有什么关系呢？ ϕ_k 应该是两 T 东西之间的关系

听到这里是否疑的，让我们继续探究

现在我们来找两个 martingale :

由刚才讲的得知: ① $\frac{S_t}{B_t} = Z_t$

weird one ② $\eta_t = \bar{E}_t^{\alpha} [\text{pay off at } T]$

① 显然是 martingale. 但②为什么是 martingale?

For η_t to be a martingale:

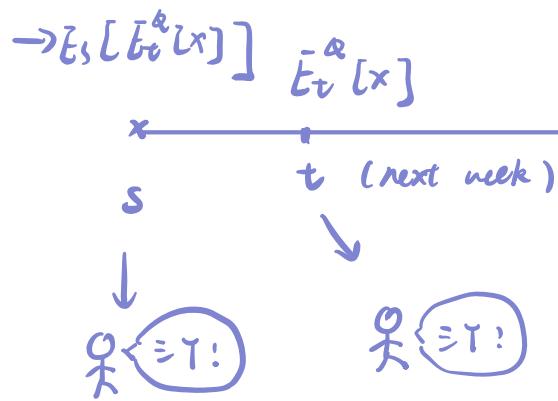
$$\bar{E}_t^{\alpha} [\eta_T] = \eta_t$$

According to the definition: set

$$\bar{E}_s^{\alpha} [\bar{E}_t^{\alpha} [x]] = \bar{E}_s^{\alpha} [x]$$

这下你套环是啥呢?

$\rightarrow \bar{E}_s [x]$



“猜猜我袋子里有几个苹果：
猜对了我什么都给你”

- tower law

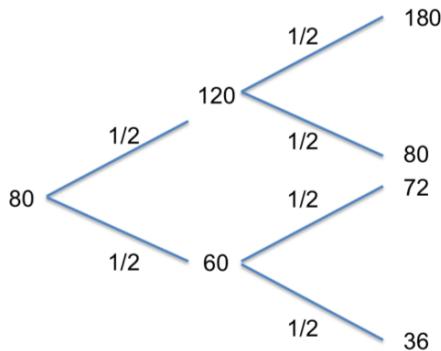
Anir: 这个 Anir 很重要

What result would ensure that η_t was a martingale ($s > t$)?

- $\eta_t = \eta_s$.
- $\eta_t = E_t[\eta_s]$ ✓ \Rightarrow 我们只管一个
- $X_t = E_t[X_s]$ \rightarrow x 的定义是 maturity

下面是一个解释的例子：（不是很重要，但如果不懂，那就很重要）

- Let's look at a stock price process:



and now consider

$$\eta_t = E_t[S_2]$$

for example.

- And so

$$\eta_0 = \frac{180 + 80 + 72 + 36}{4} = 92$$

- In the up state

$$\eta_1 = \frac{180 + 80}{2} = 130$$

and in the down state

$$\eta_1 = \frac{72 + 36}{2} = 54$$

- And so let's analyze η_t ,

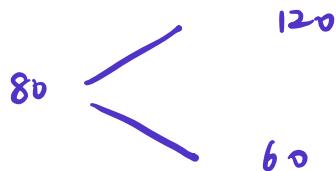
显然公式都是一样的

$$E_0[\eta_1] = \frac{130 + 54}{2} = 92$$

which is of course, η_0 and so η_t is trivially a martingale. 就不是martingale吗？

一句话在 $t=0$ 的时间点上，不管是到 t_1 还是 t_2 ，其期望都是 92

一看起来很简单，但是有一个核心问题：



如果只看这一节，那 $E_0[\eta_1]$ 应该是 $\frac{1}{2} \times (120 + 60) = 90$ 。

为什么不是呢？因为 η_1 是对 $t=1$ 时的期望！

这例子有什么问题呢： $E_t[x_1] \neq x_t$

下面我们将再回到 ①, ② 的 martingale 中, 根据我们的定理.

$$\begin{aligned} \gamma_t &= \gamma_0 + \sum_{k=1}^t \phi_k \Delta Z_k & Z_t &= \frac{S_t}{B_t} \rightarrow \text{且是 } \max(s-k, 0) \\ (\Delta \gamma_t = \phi_t \Delta Z_t) & & \text{由于 } X \text{ 的传递性, 我们取 } X = \frac{X}{B_T} \\ \gamma_t &= E^\alpha_t \left[\frac{X}{B_T} \right] \end{aligned}$$

然后我们来构建一个 portfolio: 这是在干嘛呢? 像二叉树一样, 这是 replicate option at time t , 我们买 ϕ_{t+1} units of stock and payoff at T

Q 为什么是 $t+1$, t 又能如何 $\psi_{t+1} = \gamma_t - \phi_{t+1} \frac{S_t}{B_t}$ of cash bond us 及 $t+1$ 是宏观的不断调整我们的 ϕ 使最终的 payoff 相当于我们 portfolio 的 payoff

$$\begin{aligned} \Pi_0 &= \phi_1 S_0 + \psi_1 B_0 = \phi_1 S_0 + \left(\gamma_0 - \phi_1 \frac{S_0}{B_0} \right) B_0 = \gamma_0 B_0 & = E^\alpha_0 [X] \cdot \frac{1}{B_0 e^{rT}} \cdot B_0 \\ \text{易于理解:} & & = E^\alpha_0 [X] \cdot e^{-rT} \end{aligned}$$

在 $t=1$ 时这个 portfolio 值多少钱呢? (按这个 portfolio 的面值)

$$\Pi_1 = \phi_2 S_1 + \psi_2 B_1 = \phi_2 S_1 + \left(\gamma_1 - \phi_2 \frac{S_1}{B_1} \right) B_1 = \gamma_1 B_1 \quad \begin{array}{l} \text{Self-financing!} \\ \Delta B \uparrow \\ \Delta \gamma = \phi \Delta Z \end{array}$$

(一开始的 portfolio 从 $S_0 \rightarrow S_1$, $B_0 \rightarrow B_1$ 会如何呢?)

$$\Pi_1 = \phi_1 S_1 + \psi_1 B_1 = \phi_1 S_1 + \left(\gamma_0 - \phi_1 \frac{S_0}{B_0} \right) B_1 = \gamma_0 B_1 + B_1 \phi_1 \left(\frac{S_1}{B_1} - \frac{S_0}{B_0} \right) = \gamma_0 B_1 + B_1 \Delta \gamma = B_1 \gamma_1$$

我们已经发现规律了, 那么 $\Pi_2 = \gamma_2 B_2$, $\Pi_3 = \gamma_3 B_3$, ..., $\Pi_T = \gamma_T B_T = X$
 $(E^\alpha_T \left[\frac{X}{B_T} \right] = \frac{X}{B_T})$

在 no-arbitrage 的情况下 (上面的 T 人理解是正确的情况下):

$$V_0 = \Pi_0 = B_0 \gamma_0 = B_0 E^\alpha_0 \left[\frac{\text{payoff}}{B_T} \right]$$

如果 $B_T = B_0 e^{rT}$ (As usual), 那么 $V_0 = e^{-rT} E^\alpha_0 [\text{payoff}]$

↑
这就是我们一直在用的公式呀!

Q 一个核心问题是: 中里哪来的呢? "previsible" → 怎么看的? S_t , B_t 都可见
 这类似于 self-financing: 我们 t_0 的 portfolio 到了 t_1 , 不需要加钱就可以改换成
 t_1 的 portfolio

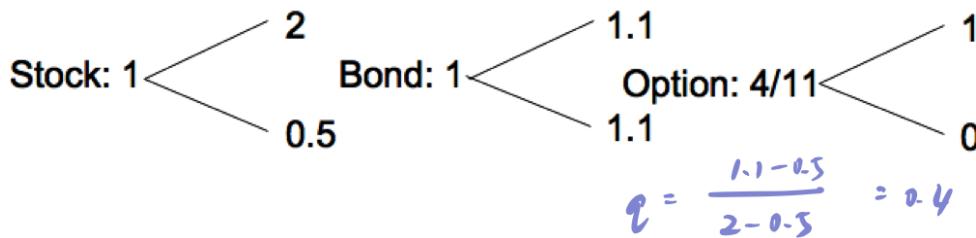
Quiz: What's true about the process $\frac{V_t}{B_t}$ here?

- It has a positive drift
- It is a martingale
- It mean reverts

$$\begin{aligned} \frac{V_t}{B_t} \text{ 里啥呢? } V_t &= B_t E^{\alpha} \left[\frac{\xrightarrow{\text{payoff}} V_T}{B_T} \right] \\ &= B_t \gamma_t \rightarrow \frac{V_t}{B_t} \text{ 是 T martingale} \end{aligned}$$

-T奇怪的问题：我们选择的 γ 使 $\frac{S}{B}$ 是 martingale，为啥不让 S 是 martingale

Consider a stock, bond and option in a one period world



$$\text{在这 T 没设置下, } E_0^{\alpha} \left[\frac{S_1}{B_1} \right] = \frac{2 \times 0.4 + 0.5 \times 0.6}{1.1} = 1 = \frac{1}{1} = \frac{S_0}{B_0}$$

$$E_0^{\alpha} \left[\frac{V_1}{B_1} \right] = \frac{1 \times 0.4 + 0 \times 0.6}{1.1} = \frac{0.4}{1.1} = \frac{V_0}{B_0}$$

但 q 不一定是 0.4 呀？我们不能让 S 是 martingale 呀？

Quiz 2: What choice of q and $1 - q$ would make the **stock price** a martingale?

- It's not possible
- $q = 2/3, 1 - q = 1/3$
- $q = 0.5, 1 - q = 0.5$
- $q = 1/3, 1 - q = 2/3$

$$2q + 0.5(1-q) = 1$$

$$1 - 5q = 0.5 \Rightarrow q = \frac{1}{3}$$

(展示了把 S 变成 martingale 是可行的)，因为 q 是什么是由我们定的

- 所以会出现什么结果呢？

q 是 $\frac{1}{3}$ 了。 S 是 martingale 了。那么 \mathbb{Q} 下的两个 martingale 变成了 γ_t 和 s_t

$$\gamma_t = \gamma_0 + \sum_{k=1}^t \phi_k \circ s_k$$

所以，我们试图搞一个 replicating portfolio 出来：

$t=0$: ϕ_1 of stock

$\psi_1 = \gamma_0 - \phi_1 s_0$ in the risk-free asset

$$\begin{aligned} \text{Then } \Pi_1 &= \phi_1 s_1 + \psi_1 B_1 \\ &= \phi_1 s_1 + (\gamma_0 - \phi_1 s_0) B_1 \\ &= \phi_1 (s_1 - s_0 B_1) + \gamma_0 B_1 \end{aligned}$$

- 此前我们能搞出来一个 self-financing portfolio，现在就不行了

S/B is a martingale

$$\Pi_0 = \gamma_0 B_0$$



$$t=1 \quad \Pi_1^{\text{old}} = \gamma_1 B_1$$

$$\Pi_1^{\text{new}} = \gamma_2 B_2$$



⋮

S is a martingale

$$\Pi_0 = \gamma_0 (= \gamma_0 B_0)$$



$$\Pi_1 = \phi_1 (s_1 - s_0 B_1) + \gamma_0 B_1$$

$$\neq \gamma_1 B_1$$

?

以上是 discrete time 的情形，下面我们讲一下 continuous time 的情形

和 discrete 的方法差不多：先找 \bar{M}_t . 然后找 replicating portfolio. 再定价
让其是 martingale

同样的 $E_s[M_t] = M_s$

$E[|M_t|] < \infty$ (在 Finance 中我们可以不考虑这)

在 Continuous time 中有一下现成的 martingale :

If $w(t)$ is a Brownian Motion, then it is a martingale
为啥呢？

① $W(T) \xrightarrow{\text{从 } x(t) \text{ 变了马甲}} = w(t) + \int_t^T dw(s)$

$$E_t[W(T)] = w(t) + 0 = w(t)$$

还有另一下和 discrete 同样的 martingale

② $\gamma_t = E_t^P[x]$

因为 $E_s^P[E_t^P[x]] = E_s^P[x]$ (tower law)

轻松一下。一下没什么用的 Quiz :

Which of the following are martingales?

- Stock price, S
- Option price, V
- Bond price, B
- $\sigma S dW$ ✓
- Stock price, S under risk-neutral measure Q
- Stock price, S under risk-neutral measure Q when R = 0 ✓
- Bond price, B under risk-neutral measure Q.

在 Continuous 中也有两个 martingales 的关系：

$$N_t = N_0 + \int_0^t \phi_s dM_s \quad * \phi \text{ is unique}$$

这 T 关系为我们提供了便利 一 为 呢？

在 continuous time 的世界里，stochastic process 多种多样：

$$ds = adt + Sdw$$

$$ds = \delta^2 e^{S^2} dw$$

$$ds = (a - bs)dt + Sdw$$

上面这 T 关系告诉我们一 T 非常实用的定理：

 In general we find that if X is a stochastic process with volatility σ_t , $dX_t = \mu_t dt + \sigma_t dW_t$ then

$$X \text{ is a martingale} \Leftrightarrow X \text{ is driftless, } \underline{\mu_t = 0}$$

→ 此项没有系数

一 简单粗暴：有 dt 就不行

休息时间：应用这 T 定理的 quiz：

Which of the following are martingales?

- S where $dS = \mu S dt + \sigma S dW$.
- B where $dB = rB dt$.
- F where $dF = \sigma F dW$ ✓
- V where $dV = rdt + \sigma dW$
- V/B where $d(V/B) = \sigma(V/B)dW$ ✓
- r where $dr = k(\theta - r)dt + \sigma dW$

diffusion term 爱啥啥舍

注意：这里有一个隐藏的，但也是非常核心的结论：

If X_t is a martingale then it can be written as:

$$X_t = X_0 + \int_0^t \phi_s dW_s$$

where W is a Brownian motion and so $dX_t = \phi_t dW_t$.

这里在说什么呢？任何 martingale 的变化都可以用一个系数乘 dW 来表示

下面是非常困难的搞出 replicating portfolio 的过程：

$$\Delta \Pi_t = \phi_t \Delta S_t + \psi_t \Delta B_t$$

目的是 replicate payoff at T . 那么有：

$$\Pi_T = \phi_T S_T + \psi_T B_T = X$$

这里有一个困难：在 discrete 中，我们只需 replicate portfolio，找出 q . ($\frac{S}{B}$ 是干嘛)

但 continuous time 中， q 是哪来的？（因为连续了）

下面到了 Girsanov's Theorem 的表演时间：

★ If W is a Brownian motion under the probability measure P , then there exists an equivalent measure Q such that $\tilde{W}_t = W_t + \gamma t$ is a Brownian motion

注意：此时我们不管 P 是怎么到 Q 的，只需知道我们能这么做

— “为什么你把你的房子烧着了？” “因为，我熊”

上面的 P 看作 Real - World Probability, 而不是另一个世界, 有一套概率分布

$$\text{under } P : dx = \mu_t dt + \sigma_t d\tilde{w}$$

\hookrightarrow Brownian Motion

$$\begin{aligned} \text{under } Q : dx &= \mu_t dt + \sigma_t (d\tilde{w} - \gamma dt) \\ &= (\underline{\mu_t - \gamma \sigma_t}) dt + \sigma_t d\tilde{w} \\ &\text{new drift, 取名 } v_t \\ \text{需要满足 } \gamma &= \frac{\mu_t - v_t}{\sigma_t} \end{aligned}$$

我们想干嘛? 在 discrete time 中我们让 q 满足 $\frac{S_t}{B_t}$ is a martingale
在 continuous time 中我们让 Q 满足 \underline{x} is a martingale
 x 是啥一会儿说

$$\text{怎么做呢? } v_t = 0 \Rightarrow \gamma = \frac{\mu_t}{\sigma_t}$$

下面我们就找 replicating portfolio (ϕ_t, ψ_t)

continuous 我们也让 $\frac{S_t}{B_t}$ 是 martingale

$$ds = \mu_s dt + \sigma_s d\tilde{w}$$

$$dB = rB dt$$

$$d\left(\frac{S}{B}\right) = ds \cdot \frac{1}{B} + s \cdot d\left(\frac{1}{B}\right) + \underbrace{ds d\left(\frac{1}{B}\right)}_0$$

困扰我们的是 $d\left(\frac{1}{B}\right)$:

用 Ito's Lemma $\gamma = \frac{1}{B}$

$$dY = \frac{dY}{dB} dB + \frac{1}{2} \frac{d^2 Y}{dB^2} dB^2 = -\frac{1}{B^2} dB + \frac{1}{2} \cdot \frac{2}{B^3} dB^2 = -\frac{r}{B} dt$$

$$\text{原式} = (\mu s dt + \sigma s dW) \frac{1}{B} - r \cdot \frac{s}{B} dt$$

$$= (\mu - r) \frac{s}{B} dt + \sigma \frac{s}{B} dW$$

$\downarrow z = \frac{s}{B}$

$$dz = (\mu - r) z dt + \sigma z dW$$

不忘初心，牢记使命：我们是想让 S 是 martingale，找能让它成立的 prob. measure &

我们改换门庭，让 drift 消失：

$$dz = (\mu - r) z dt + \sigma z (d\tilde{W} - \gamma dt)$$

$$\Rightarrow \gamma = \frac{\mu - r}{\sigma}, \text{ 使 } dz = \sigma z d\tilde{W}$$

这是 Sharpe ratio \rightarrow market price of risk · 把这叫做 risk premium

Qn: stock 们 follow what process?

$$ds = \mu s dt + \sigma s (d\tilde{W} - \gamma dt)$$

$$= [\mu s - (\mu - r)s] dt + \sigma s d\tilde{W}$$

$$= \underline{rs} dt + \sigma s d\tilde{W}$$

为什么是 rs 呢？如果 $\frac{s}{B}$ 是 martingale，我们可以认为这是 risk-neutral world. 此时 S 的 drift 是 rs

那么到底如何搞出来 Replicating Portfolio η_t ?

有两个 martingale: $\begin{cases} \eta_t = E_t^\alpha \left[\frac{x}{B_T} \right] \\ Z_t = \frac{s_t}{B_t} \end{cases}$

market rep. theorem $d\eta_t = \phi_t dz_t$

Portfolio 还是原来的两部分： ϕ_t of stock

$$\psi_t = \gamma_t - \phi_t z_t \text{ of bond}$$

注意！和 Binomial 有什么区别呢？系数不是 $t+1$ 了，因为 continuous time 没有 $t+1$

下面就是紧张刺激的推导过程：

$$\begin{aligned}\Pi_t &= \phi_t S_t + \psi_t B_t & d\gamma_t &= \phi_t dz_t = \phi_t \sigma z_t d\tilde{w} \\ &= \phi_t S_t + (\gamma_t - \phi_t \frac{S_t}{B_t}) B_t & dB_t &= r B_t dt \\ &= \gamma_t B_t & \text{(和 Binomial 结果是一样的)}\end{aligned}$$

接下来我们检查一下这个 portfolio 是否是 self-financing \Leftrightarrow changes in Π only come from changes in S and B

$$d\Pi_t = B_t d\gamma_t + \gamma_t dB_t + d\gamma_t dB_t \quad (\text{先从上式直接 } L\hat{w})$$

$$= B_t \phi_t dz_t + \gamma_t dB_t \quad \psi_t = \gamma_t - \phi_t z_t$$

$$= \phi_t B_t dz_t + (\psi_t + \phi_t z_t) dB_t$$

$$= \phi_t (B_t dz_t + z_t dB_t) + \psi_t dB_t$$

$$S_t = B_t \times z_t \Rightarrow dS_t = B_t dz_t + z_t dB_t + \underline{dB_t dz_t}_0$$

$$= \phi_t dS_t + \psi_t dB_t$$

这样就有 self-financing 了！ ☺

At $t=0$ $\Pi_0 = \gamma_0 B_0$ or at t $\Pi_t = \gamma_t B_t$. This grows over time in a self-financing way until at maturity T : $\Pi_T = \gamma_T B_T = \underline{x} \rightarrow \text{payoff}$

在 No-Arbitrage 的情况下.

$$V_t = \bar{V}_t = B_t e^{\alpha t} = B_t E_t^\alpha \left[\frac{\text{payoff}}{B_T} \right] \quad \text{If } B_t = B_0 e^{rt}, \text{ then}$$

$$V_t = e^{-r(T-t)} \bar{E}_t^\alpha [\text{payoff}] \quad \text{和 discrete 情况是一样的}$$

这 \bar{E}_t^α 就可以看做是 RN

所以这到底有什么用? 我们本来的难点就是算 $E_t^\alpha [\text{payoff}]$
现在这问题并没有解决.

下面我们来算 $E_t^\alpha [\text{payoff}]$

有一些方法: • Analytically \rightarrow 算出公式

• Monte Carlo \rightarrow 下周要讲

• 用 PDE

我们先来试试 Valuing a Call Option:

Under α : $V_0 = e^{-rT} E_0^\alpha [\text{payoff}] \Rightarrow$ Thus $\frac{V_t}{B_t}$ 也是 martingale

$$dS = rSdt + \delta Sd\tilde{w}$$

To value a call option: payoff = $\max(S_T - K, 0)$

通过 SDZ, 我们知道 [Stochastic Calculus & Ito's Lemma . pg 8]

$$S_T = S_0 \exp((r - \frac{1}{2}\sigma^2)T + \delta w(T))$$

$$w(T) = \sqrt{T} X \quad X \sim N(0, 1)$$

我们把 S_T 算出来，代入 payoff，求 V_0

$$V_0 = e^{-rT} E_0^{\alpha} \left[\max(S_0 e^{rT - \frac{1}{2}\sigma^2 T + \delta\sqrt{T} X} - k, 0) \right]$$

random normally distributed

下面我们要算 Expectation。我们唯一要做的期望就是 X 的期望，剩下的都是定值

$$V_0 = e^{-rT} \int_{-\infty}^{\infty} \frac{e^{-\frac{x^2}{2}}}{\sqrt{2\pi}} \times \max(S_0 e^{rT - \frac{1}{2}\sigma^2 T + \delta\sqrt{T} X} - k, 0) dx$$

↓ 很烦 \Rightarrow 把它忘了！
 prob. \Rightarrow 为什么这样呢？因为期望是 x 的所有概率 \times payoff. $x \sim N(0)$

接下来我们要 choose a value of x below when payoff is 0
 这想法非常关键！

只需把正值积起来就可以了：

$$\text{找到 } x, \text{ 使 } S_0 e^{rT - \frac{1}{2}\sigma^2 T + \delta\sqrt{T} X} > k$$

$$rT - \frac{1}{2}\sigma^2 T + \delta\sqrt{T} X > \ln \frac{k}{S_0}$$

$$\delta\sqrt{T} X > \ln \frac{k}{S_0} - (r - \frac{1}{2}\sigma^2) T$$

$$X > \underbrace{\frac{\ln \frac{k}{S_0} - (r - \frac{1}{2}\sigma^2) T}{\delta\sqrt{T}}}_{L} \rightarrow \text{and } d_1, d_2 \text{ 长得很像}$$

$$V_0 = e^{-rT} \int_L^{\infty} \frac{e^{-\frac{x^2}{2}}}{\sqrt{2\pi}} (S_0 e^{rT - \frac{1}{2}\sigma^2 T + \delta\sqrt{T} X} - k) dx$$

$$= A - B$$

$$= \underbrace{e^{-rT}}_{\text{常数}} \int_L^{\infty} \frac{e^{-\frac{x^2}{2}}}{\sqrt{2\pi}} S_0 e^{rT - \frac{1}{2}\sigma^2 T + \delta\sqrt{T} X} dx - e^{-rT} \int_L^{\infty} \frac{e^{-\frac{x^2}{2}}}{\sqrt{2\pi}} \cdot k \cdot dx$$

$$= \frac{S_0}{\delta\sqrt{\pi}} \int_L^\infty e^{-\frac{x^2}{2} - \frac{1}{2}\delta^2 T + \delta\bar{r}T} dx - Ke^{-rT} \int_L^\infty \frac{e^{-\frac{x^2}{2}}}{\delta\sqrt{\pi}} dx$$

里T无年方数：

$$= \frac{S_0}{\delta\sqrt{\pi}} \int_L^\infty e^{-\frac{1}{2}(x-\delta\bar{r}T)^2} dx$$

Let $y = x - \delta\bar{r}T$ $dy = dx$

 $x = L \quad y = L - \delta\bar{r}T$

$$= S_0 \underbrace{\int_{L-\delta\bar{r}T}^\infty \frac{1}{\delta\sqrt{\pi}} e^{-\frac{1}{2}y^2} dy}_{\text{Normal Density } \odot \textcircled{1}}$$

Recall $N(x) = \int_{-\infty}^x \frac{1}{\sqrt{\pi}} e^{-\frac{y^2}{2}} dy$ $\int_L^\infty \frac{e^{-\frac{x^2}{2}}}{\delta\sqrt{\pi}} dx = N(L)$

$$\textcircled{1}: \int_{L-\delta\bar{r}T}^\infty \rightarrow 1 - N(L - \delta\bar{r}T) = N(-(L - \delta\bar{r}T)) = N(-L + \delta\bar{r}T) = N(d_1)$$

$$= N\left(\frac{\ln \frac{S_0}{K} + (r - \frac{1}{2}\delta^2)T}{\delta\bar{r}T} + \delta\bar{r}T\right) = N(d_1)$$

$$= N(d_1)$$

$$A = S_0 N(d_1)$$

$$B = K e^{-rT} N(d_2)$$

$$= S_0 N(d_1) - K e^{-rT} N(d_2)$$

威嘴！ \odot

这方法比从 PDF 积分算出 B-S formula 简单多了

用 Martingale 还可以干嘛呢？还可以把 B-S 的 PDF 求出来！

We do some Ito's Lemma on V and B now:

under Ω :

$$dS = rSdt + \sigma S d\tilde{W}$$

and $\frac{S_t}{B_t}$, $\frac{V_t}{B_t}$ are martingales
↳ 我们准备利用这一点

$$dB = rBdt$$

$$dV = \frac{\partial V}{\partial t} dt + \frac{\partial V}{\partial S} dS + \frac{1}{2} \frac{\partial^2 V}{\partial S^2} dS^2$$

$$= \frac{\partial V}{\partial t} dt + \frac{\partial V}{\partial S} (rSdt + \sigma S d\tilde{W}) + \frac{1}{2} \sigma^2 S^2 \frac{\partial^2 V}{\partial S^2} dt$$

$$= \left(\frac{\partial V}{\partial t} + rs \frac{\partial V}{\partial S} + \frac{1}{2} \sigma^2 S^2 \frac{\partial^2 V}{\partial S^2} \right) dt + ds \frac{\partial V}{\partial S} d\tilde{W}$$

$$d\left(\frac{V}{B}\right) = \frac{1}{B} dV + V d\left(\frac{1}{B}\right) + dV d\left(\frac{1}{B}\right)$$

$$= \frac{1}{B} \left(\frac{\partial V}{\partial t} + \frac{\partial V}{\partial S} (rSdt + \sigma S d\tilde{W}) + \frac{1}{2} \sigma^2 S^2 \frac{\partial^2 V}{\partial S^2} dt \right) - \frac{rV}{B} dt$$

$$= \frac{1}{B} \left(\frac{\partial V}{\partial t} + rs \frac{\partial V}{\partial S} + \frac{1}{2} \sigma^2 S^2 \frac{\partial^2 V}{\partial S^2} - rV \right) dt + \delta \frac{S}{B} \frac{\partial V}{\partial S} d\tilde{W}$$

* $\frac{V}{B}$ is a martingale \Rightarrow drift = 0

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

Hurray! 😊

BSM PDF!

更妙的地方是什么呢？ B_t is not special

You can choose any treated asset N_t as numeraire

↓
在 QT 下 $\frac{A_t}{N_t}$ is a martingale
At 也是数位 asset

$N_t = B_t$ is the easiest case. 但什么都行. 因为反正它们的回报率都是 μ

- In the examples we have given the risk-free bond plays the role of the Numeraire asset. It is useful because it has no volatility but could we use any asset as the numeraire asset?
- In fact we can, we just have a revised pricing formula for any numeraire asset N_t so that now:

$$\frac{V_t}{N_t} = E_t^Q \left[\frac{V_T}{N_T} \right]$$

- This is very useful for foreign exchange problems and crucial for term structure work, as we will see later in the course.

接下来我们用 Exchange Option 和 Quanto 作为例子：

Exchange Option

$$dS_1 = \mu_1 S_1 dt + \sigma_1 S_1 dW_1 \quad dW_1 dW_2 = \rho dt$$

$$dS_2 = \mu_2 S_2 dt + \sigma_2 S_2 dW_2$$

$$dB = rBdt$$

$$V(S_1, S_2, T) = \max(S_2 - S_1, 0)$$

choose S_1 as "Numeraire". so there exists a measure α where $\frac{S_1}{S_1}, \frac{S_2}{S_1}$ are martingales 能不能用 B 为 numeraire 呢？能，但是太困难了

首先我们写出 $\underline{d\frac{S_2}{S_1}} = \underbrace{(\mu_2 - \mu_1 + \sigma_1^2 - \rho\sigma_1\sigma_2)}_{\mu_3} \left(\frac{S_2}{S_1} \right) dt - \sigma_1 \left(\frac{S_2}{S_1} \right) dW_1 + \sigma_2 \left(\frac{S_2}{S_1} \right) dW_2$

等号：
 $\sigma_2 = \sqrt{\sigma_1^2 + \sigma_2^2 - 2\rho\sigma_1\sigma_2} \rightarrow \sigma_2 \frac{S_2}{S_1} dW_2$

接下来我们 move to α

$$d\frac{S_2}{S_1} = \mu_3 \left(\frac{S_2}{S_1}\right) dt + \sigma_3 \left(\frac{S_2}{S_1}\right) (d\tilde{W}_3 - \gamma dt)$$

$$\gamma = \frac{\mu_3}{\sigma_3}$$

$$\Rightarrow d\frac{S_2}{S_1} = \sigma_3 \left(\frac{S_2}{S_1}\right) d\tilde{W}_3$$

solve under α

$$\frac{S_2(T)}{S_1(T)} = \frac{S_2(0)}{S_1(0)} \exp\left(-\frac{1}{2} \sigma_3^2 T + \sigma_3 \sqrt{T} X\right) \quad X \in N(0, 1)$$

Also, $\frac{V}{S_1}$ is a martingale too.

$$\frac{V(0)}{S_1(0)} = E_0^\alpha \left[\frac{V(T)}{S_1(T)} \right]$$

$$V(0) = S_1(0) E_0^\alpha \left[\frac{\max(S_2(T) - S_1(T), 0)}{S_1(T)} \right]$$

$$= S_1(0) E_0^\alpha \left[\max \left(\underbrace{\frac{S_2(T)}{S_1(T)} - 1}_{\text{no drift}}, 0 \right) \right]$$

no drift

$$V(0) = S_1(0) \left[\frac{S_2(0)}{S_1(0)} N(d_1) - 1 \times N(d_2) \right]$$

$$\text{其中 } d_1 = \frac{\ln \frac{S_2(0)}{S_1(0)} + \frac{1}{2} \sigma_3^2 T}{\sigma_3 \sqrt{T}}$$

$$d_2 = d_1 - \sigma_3 \sqrt{T}$$

$$V(0) = S_2(0) N(d_1) - S_1(0) N(d_2)$$

当然我们可以用老办法 (Lecture 11)，和以前的做法都是一样的

Quantitative Option (最后的部分)

$$dS = \mu_S dt + \sigma_S dW_1$$

$$\left. \begin{array}{l} \\ \end{array} \right\} dW_1/dm = p(t)$$

$$de = \mu_E dt + \sigma_E dW_2 \quad (\text{exchange rate})$$

$$dB\$ = r\$ B\$ dt$$

$$dB_E = r_E B_E dt$$

Eventually $B\$$ will be numeraire

- Problem \Rightarrow need traded assets denominated in \$

$$\frac{A_t}{N_t} \Rightarrow A_t, N_t \text{ must be in the same currency}$$

We are going to introduce Y : dollar value of Eurobond

Z : dollar value of Eurostoxx index

在这种设置下，我们有：

(Quiz) $\frac{Y}{B\$}, \frac{Z}{B\$}$ will be martingales (因为) USD denominated asset $\frac{Y}{B\$}$ 都是

Now e is not a traded asset

\hookrightarrow 买了这 e exchange rate

$\frac{e}{B\$}$ 就不是 martingale 了

$$dY = e dB_E + B_E de = (\mu_E + r_E) Y dt + \sigma_E Y dW_2$$

$$dZ = S de + e ds + deds = (\mu + \mu_E + \rho \sigma_E) Z dt + \sigma_Z Z dW_1 + \sigma_E Z dW_2$$

$K = \frac{Y}{B\$}$ is a martingale

$$dK = \frac{1}{B\$} dY + d\left(\frac{1}{B\$}\right) Y$$

$$= (r_E + \mu_E - r_{\$}) K dt + \delta_E K dW_2$$

$$\gamma_2 = \frac{r_E + \mu_E - r_{\$}}{\delta_E}$$

$$\Rightarrow dK = \delta_E K d\tilde{W}_2 \text{, where } d\tilde{W}_2 = dW_2 + \gamma_2 dt$$

$L = \frac{Z}{B\$}$ is also a martingale

$$dL = \frac{1}{B\$} dZ + d\left(\frac{1}{B\$}\right) Z$$

$$= (\mu + \mu_E + \rho \delta_E \sigma - r_{\$}) L dt + \delta L dW_1 + \delta_E L dW_2$$

$$\text{But under } Q, dW_2 \rightarrow d\tilde{W}_2 - \gamma_2 dt$$

$$dL = (\mu + \mu_E + \rho \delta_E \sigma - r_{\$} - (r_E + \mu_E - r_{\$})) L dt + \delta L dW_1 + \delta_E L d\tilde{W}_2$$

$$= (\mu + \rho \delta_E \sigma - r_E) L dt + \delta L dW_1 + \delta_E L d\tilde{W}_2$$

L is a martingale

$$dW_1 \rightarrow d\tilde{W}_1 - \gamma_1 dt$$

$$\gamma_1 = \frac{\mu + \rho \delta_E \sigma - r_E}{\delta} \quad \Rightarrow dL = \delta L d\tilde{W}_1 + \delta_E L d\tilde{W}_2$$

Finally. $\frac{V}{B\$}$ is also a martingale

$$V_0 = e^{-rT} E_0^\alpha [V_T]$$

$$= e^{-rT} E_0^\alpha [\max(S_T - K, 0)]$$

We need to know what is S_T under α

But recall

$$dS = \mu S dt + \sigma S dW_1$$

So under α

$$\begin{aligned} dS &= (\mu - \rho\gamma) S dt + \sigma S d\tilde{W}_1 \\ &= (r_\epsilon - \rho\sigma_e\sigma) S dt + \sigma S d\tilde{W}_1 \end{aligned}$$

$$S_T = S_0 \exp((r_\epsilon - \rho\sigma_e\sigma - \frac{1}{2}\sigma^2)T + \sigma\sqrt{T} X) \quad X \sim N(0, 1)$$

或者

- And so, $\frac{V}{B\$} = \frac{1}{B\$} dV + V d\left(\frac{1}{B\$}\right)$, where $dV = \frac{\partial V}{\partial t} dt + \frac{\partial V}{\partial S} ds + \frac{1}{2} \frac{\partial^2 V}{\partial S^2} ds^2$

$$d\left(\frac{V}{B\$}\right) = \frac{1}{B\$} \left(\frac{\partial V}{\partial t} + (r_\epsilon - \rho\sigma_e\sigma)S \frac{\partial V}{\partial S} + \frac{1}{2}\sigma^2 S^2 \frac{\partial^2 V}{\partial S^2} - r\$ V \right) dt + \sigma \frac{S}{B\$} \frac{\partial V}{\partial S} d\tilde{W}_1$$

- But, under the risk neutral world V/B must be a martingale, in which case it must have zero drift and,

$$\frac{\partial V}{\partial t} + (r_\epsilon - \rho\sigma_e\sigma)S \frac{\partial V}{\partial S} + \frac{1}{2}\sigma^2 S^2 \frac{\partial^2 V}{\partial S^2} - r\$ V = 0$$

which, again, is the Black-Scholes-Merton PDE but now with a different first derivative term

Expectation & RN prob. 和 PDE 是一样的