

- * underlying source of uncertainty $(X_t)_{0 \leq t \leq T}$ is an Ito process and satisfies the SDE:

$$dX_t = \mu^X(t, X_t) dt + \sigma^X(t, X_t) dW_t$$

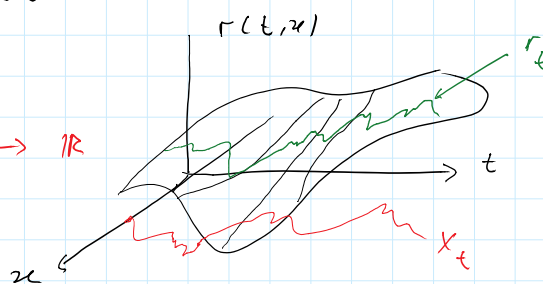
$$\mu^X(t, x) : \mathbb{R}_+ \times \mathbb{R} \rightarrow \mathbb{R} \quad \sigma^X(t, x) : \mathbb{R}_+ \times \mathbb{R} \rightarrow \mathbb{R}_+$$

X_t is not the price of a traded asset

- * Bank account $(B_t)_{0 \leq t \leq T}$ is driven by X through the short rate process $(r_t)_{0 \leq t \leq T}$

$$r_t = r(t, X_t)$$

$$r : \mathbb{R}_+ \times \mathbb{R} \rightarrow \mathbb{R}$$



B satisfies the SDE: $dB_t = r_t B_t dt$

- * Traded risky asset $(F_t)_{0 \leq t \leq T}$ and we assume it is Ito process & that F satisfies SDE

$$dF_t = F_t \mu^F(t, X_t) dt + F_t \sigma^F(t, X_t) dW_t$$

instantaneous return of asset or drift. volatility

$$\mu_t^X = \mu^X(t, X_t), \quad \sigma_t^X = \sigma^X(t, X_t)$$

$$\mu_t^F = \mu^F(t, X_t), \quad \sigma_t^F = \sigma^F(t, X_t)$$

- * Throw in a third asset $(y_t)_{0 \leq t \leq T}$ and pays $G(X_T)$ at maturity T .

What properties must g have to avoid arbitrage?

- set up a self-financing strategy $(\alpha_t, \beta_t)_{0 \leq t \leq T}$ and show 1 of g . Let $(V_t)_{0 \leq t \leq T}$ be the value of the strategy.
 - \uparrow # of F
 - \uparrow # of B

$$V_t = \alpha_t F_t + \beta_t B_t - g_t$$

- specifically set $V_0 = 0$.

$$\begin{aligned} x \quad dV_t &= d(\alpha_t F_t) + d(\beta_t B_t) - dg_t \\ &= \underbrace{d\alpha_t F_t + \alpha_t dF_t + d[\alpha, F]_t}_{\text{self-financing condition}} \\ &\quad + \underbrace{d\beta_t B_t + \beta_t dB_t + d[\beta, B]_t} - dg_t \\ &= \underbrace{\alpha_t}_{\text{self-financing condition}} dF_t + \underbrace{\beta_t}_{\text{self-financing condition}} dB_t - \underbrace{dg_t}_{\text{self-financing condition}} \end{aligned}$$

let's assume that $g_t = g(t, X_t)$ and $g \in C^{1,2}$ then \leftarrow what is this function g .

$$dg_t = g_t u^g(t, X_t) dt + g_t \sigma^g(t, X_t) dW_t$$

$$\begin{aligned} \Rightarrow dV_t &= \alpha_t (F_t u^f dt + F_t \sigma^f dW_t) + \beta_t r_t B_t dt \\ &\quad - (g_t u^g dt + g_t \sigma^g dW_t) \\ &= (\alpha_t F_t u^f - g_t u^g + \beta_t r_t B_t) dt \\ &\quad + (\alpha_t F_t \sigma^f - g_t \sigma^g) dW_t \end{aligned}$$

locally remove risk:

$$\alpha_t = \frac{g_t \sigma_t^g}{F_t \sigma_t^f}$$

now: $dV_t = A_t dt + A_t t F_t$

\Rightarrow to avoid arbitrage, we must have $A_t = 0 \forall (t, X_t)$

$$\Rightarrow dV_t = 0 \quad \forall t$$

$$\Rightarrow V_t = 0 \quad \forall t$$

$$\Rightarrow \alpha_t F_t + \beta_t B_t - g_t = 0$$

$$\Rightarrow \boxed{\beta_t B_t = g_t - \alpha_t F_t}$$

$$A_t = 0 \Leftrightarrow \alpha_t F_t + \mu_t^f + r_t \beta_t B_t = \mu_t^g g_t$$

$$\Leftrightarrow \mu_t^f \frac{\sigma_t^g}{\sigma_t^f} g_t + r_t (g_t - \frac{\sigma_t^g}{\sigma_t^f} g_t) = \mu_t^g g_t$$

$$\Leftrightarrow \boxed{\frac{\mu_t^f - r_t}{\sigma_t^f} = \frac{\mu_t^g - r_t}{\sigma_t^g}} = \lambda_t = \lambda(t, X_t)$$

"Sharpe ratio"

has to be the same
(λ_t) over all tradable assets!
"Market price of risk"

$$\frac{\mu_t^g - r_t}{\sigma_t^g} = \lambda_t \Leftrightarrow \mu_t^g - \lambda_t \sigma_t^g = r_t$$

recall Itô's lemma says that $g_t = g(t, X_t)$ satisfies the SDE

$$dg_t = \left[\partial_t g(t, X_t) + \mu_t^x(t, X_t) \partial_x g(t, X_t) + \frac{1}{2} (\sigma_t^x(t, X_t))^2 \partial_{xx} g(t, X_t) \right] dt + \sigma_t^x(t, X_t) \partial_x g(t, X_t) dW_t$$

$$\begin{aligned} \partial_t g_t &:= \partial_t g(t, X_t) \\ \partial_x g_t &:= \partial_x g(t, X_t) \end{aligned}$$

$$dg_t = \left(\partial_t g_t + \mu_t^x \partial_x g_t + \frac{1}{2} (\sigma_t^x)^2 \partial_{xx} g_t \right) dt + (\sigma_t^x \partial_x g_t) dW_t$$

$g_t \mu_t^g$

$\sim g_t \sigma_t^g$

$$\mu_t^g - \lambda_t \sigma_t^g = r_t \Rightarrow \mu_t^g g_t - \lambda_t \sigma_t^g g_t = r_t g_t$$

$$\Rightarrow \partial_t g_t + \mu_t^x \partial_x g_t + \frac{1}{2} (\sigma_t^x)^2 \partial_{xx} g_t - \lambda_t \sigma_t^x \partial_x g_t = r_t g_t$$

$$\Rightarrow \partial_t g_t + \mu_t^X \partial_x g_t + \frac{1}{2} (\sigma_t^X)^2 \partial_{xx} g_t - \lambda_t \sigma_t^X \partial_x g_t = r_t g_t$$

$$\Rightarrow \partial_t g_t + (\mu_t^X - \lambda_t \sigma_t^X) \partial_x g_t + \frac{1}{2} (\sigma_t^X)^2 \partial_{xx} g_t = r_t g_t$$

holds $\forall (t, x) \in [0, T] \times \mathbb{R}$ and so the function g satisfies the PDE:

$$\partial_t g(t, x) + (\mu^X(t, x) - \lambda(t, x) \sigma^X(t, x)) \partial_x g(t, x) + \frac{1}{2} (\sigma^X(t, x))^2 \partial_{xx} g(t, x) = r(t, x) g(t, x)$$

terminal condition $g(T, x) = G(x)$

generalized Black-Scholes PDE.

If X_t is traded, e.g. by choosing $F_t = X_t$

$$\Rightarrow df_t = \mu_t^X dt + \sigma_t^X dW_t$$

$$= f_t \mu_t^f dt + f_t \sigma_t^f dW_t$$

$$= X_t \mu_t^f dt + X_t \sigma_t^f dW_t$$

$$\Rightarrow \mu_t^f = \mu_t^X / X_t, \quad \sigma_t^f = \sigma_t^X / X_t$$

$$\frac{\mu_t^f - r_t}{\sigma_t^f} = \lambda_t \Rightarrow \mu_t^f - \lambda_t \sigma_t^f = r_t$$

$$\mu_t^X - \lambda_t \sigma_t^X = r_t X_t$$

when X is traded we have:

$$\left\{ \begin{array}{l} \partial_t g + r x \partial_x g + \frac{1}{2} (\sigma^X)^2 \partial_{xx} g = r g \\ g(T, x) = G(x) \end{array} \right.$$

try repeating with X traded from the start.

suppose that X is traded and $\frac{dX_t}{X_t} = \mu dt + \sigma dW_t$

suppose that X is traded and $\frac{dX_t}{X_t} = \mu dt + \sigma dW_t$

(i.e. $dX_t = \underbrace{X_t \mu}_{\mu^X} dt + \underbrace{\sigma X_t}_{\sigma^X} dW_t$) (Black-Scholes model)

$$\Rightarrow \begin{cases} \partial_t g + r x \partial_x g + \frac{1}{2} \sigma^2 x^2 \partial_{xx} g = r g, \\ g(t, x) = G(x) \end{cases}$$

Black-Scholes PDE.

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$$\begin{cases} \partial_t g + rx \partial_x g + \frac{1}{2} \sigma^2 x^2 \partial_{xx} g = rg, \\ g(T, x) = 1. \end{cases}$$

$r(t, x) = r(t)$ only a fn. of time

this a bond with deterministic discounting, so expect that

$$g(t, x) = g(t) \text{ only a fn of } t)$$

$$\partial_t g(t) = r(t) g(t), \quad g(T) = 1$$

$$\Rightarrow \frac{\partial_t g(t)}{g(t)} = r(t) \Rightarrow \partial_t \log g(t) = r(t)$$

$$\Rightarrow \int_t^T \partial_t \log g(u) du = \int_t^T r(u) du$$

$$\Rightarrow \log g(T) - \log g(t) = \int_t^T r(u) du$$

$$\Rightarrow g(t) = e^{-\int_t^T r(u) du}$$

$$\begin{cases} \partial_t g + rx \partial_x g + \frac{1}{2} \sigma^2 x^2 \partial_{xx} g = rg, \\ g(T, x) = x \end{cases}$$

this is just a claim that pays X_T at T , so it is no different from the asset itself.

$$g(t, x) = x$$

$$l.h.s. = 0 + rx - 1 + \frac{1}{2} \sigma^2 x^2 \cdot 0 \quad \checkmark$$

$$r.h.s. = rx$$

terminal condition $g(T, x) = x \quad \checkmark$

$$\begin{cases} \partial_t g + rx \partial_x g + \frac{1}{2} \sigma^2 x^2 \partial_{xx} g = rg, \\ g(T, x) = x^2 \end{cases}$$

$$\dots \dots \dots 2(r - \frac{1}{2} \sigma^2) T + 2\sigma^2 T^2 \dots$$

l

$$g(t, x) = x^2$$

$$\begin{aligned} E_0^Q [X_T^2] &= E_0^Q [X_0^2 e^{2(r - \frac{1}{2}\sigma^2)T + 2\sigma Z}] \\ &= X_0^2 E_0^Q [e^{2(r - \frac{1}{2}\sigma^2)T + 2\sigma Z}] \\ &= X_0^2 E_0^Q [1] \end{aligned}$$

$Z \sim N(0, 1)$

perhaps $g(t, x) = l(t) x^2$

$$g(t, x) = x^2 \Rightarrow l(T) = 1$$

$$\Rightarrow x^2 \partial_t l + r x \cdot 2x \cdot l + \frac{1}{2} \sigma^2 x^2 \cdot 2l = r l x^2$$

$$\partial_t l + (2r + \sigma^2) l = r l$$

$$\partial_t l + (r + \sigma^2) l = 0$$

$$\Rightarrow l(t) = e^{-(r + \sigma^2)(T - t)}$$

$$g(t, x) = e^{-(r + \sigma^2)(T - t)} x^2$$

try $g(x) = x^2$

solving parabolic PDEs.

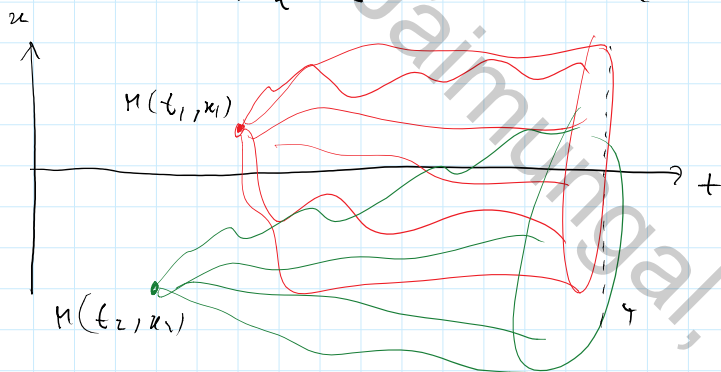
$$\begin{cases} \partial_t h(t, x) + \frac{1}{2} \partial_{xx} h(t, x) = 0, \\ h(T, x) = H(x). \end{cases}$$

Feynman-Kac Theorem:

Suppose h satisfies the above, then it admits the following representation:

$$h(t, x) = \mathbb{E}[H(X_T) \mid X_t = x]$$

X_t is a Brownian motion.



$$g_t \triangleq \mathbb{E}[H(X_T) \mid \mathcal{F}_t], \quad \mathcal{F} \text{ is the natural filtration generated by } X.$$

g_t is in fact a martingale (Doob - necessary also)

let $s < t$ and check:

$$\begin{aligned} \mathbb{E}[g_t \mid \mathcal{F}_s] &= \mathbb{E}[\mathbb{E}[H(X_T) \mid \mathcal{F}_t] \mid \mathcal{F}_s] \\ &= \mathbb{E}[H(X_T) \mid \mathcal{F}_s] = g_s \end{aligned}$$

law of iterated expectation

make $g_t = \mathbb{E}[H(X_T) | \mathcal{F}_t]$

$$= \mathbb{E}[H(\underbrace{(X_T - X_t)} + \underbrace{X_t}) | \mathcal{F}_t]$$

$$= \mathbb{E}[H((T-t)^{1/2} z + X_t) | \mathcal{F}_t]$$

$$= g(t, X_t) \in C^{1,2} \quad z \sim N(0,1) \perp X_t$$

For sufficiently well behaved H

Ito For B.M.

$$\Rightarrow dg = \left(\partial_t g(t, X_t) + \frac{1}{2} \partial_{xx} g(t, X_t) \right) dt + \partial_x g(t, X_t) dW_t$$

$$\Rightarrow \int_t^{t+h} \dots = \int_t^{t+h} \dots \quad (h > 0)$$

$$g(t+h, X_{t+h}) - g(t, X_t) = \int_t^{t+h} \left(\partial_t g(u, X_u) + \frac{1}{2} \partial_{xx} g(u, X_u) \right) du + \int_t^{t+h} \partial_x g(u, X_u) dW_u$$

$$\mathbb{E}[\cdot | X_t = x]$$

$$\mathbb{E}[g_{t+h} - g_t | X_t = x] = \mathbb{E}\left[\int_t^{t+h} \left(\partial_t + \frac{1}{2} \partial_{xx} \right) g(u, X_u) du \mid X_t = x \right] + \mathbb{E}\left[\int_t^{t+h} \partial_x g(u, X_u) dW_u \mid X_t = x \right]$$

$\hookrightarrow 0$
by martingale prop.

$\hookrightarrow 0$ As Ito integrals are 0 mean.

$$\begin{aligned} 0 &= \lim_{h \downarrow 0} \mathbb{E}\left[\frac{1}{h} \int_t^{t+h} \left(\partial_t + \frac{1}{2} \partial_{xx} \right) g(u, X_u) du \mid X_t = x \right] \\ &= \mathbb{E}\left[\left(\partial_t + \frac{1}{2} \partial_{xx} \right) g(t, X_t) \mid X_t = x \right] \\ &= \left(\partial_t + \frac{1}{2} \partial_{xx} \right) g(t, x) \end{aligned}$$

i.o. $\partial_t g(t, x) + \frac{1}{2} \partial_{xx} g(t, x) = 0$

and we also have that $g_T = g(T, x) = \mathbb{E}[H(X_T) | X_T = x]$

$$= H(z)$$

so $g(z)$ satisfies the required VDE!

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$$\begin{cases} \partial_t h(t, x) + \underbrace{a(t, x)} \partial_x h + \frac{1}{2} \underbrace{b^2(t, x)} \partial_{xx} h(t, x) = \underbrace{c(t, x)} h(t, x), \\ h(T, x) = h(x). \end{cases}$$

fixed & given.

Feynman-Kac Theorem:

$$h(t, x) = \mathbb{E}^{\mathbb{P}^*} \left[e^{-\int_t^T \underbrace{c(s, X_s)} ds} H(X_T) \mid X_t = x \right]$$

where X_t satisfies the SDE:

$$dX_t = \underbrace{a(t, X_t)} dt + \underbrace{b(t, X_t)} dW_t^*$$

W_t^* is a \mathbb{P}^* -B.mtn.

$$\begin{aligned} \partial_t g(t, x) + \underbrace{(u^x(t, x) - \lambda(t, x) \sigma^x(t, x))} \partial_x g(t, x) \\ + \frac{1}{2} \underbrace{(\sigma^x(t, x))^2} \partial_{xx} g(t, x) = \underbrace{r(t, x)} g(t, x) \end{aligned}$$

terminal condition $g(T, x) = G(x)$

$$\Rightarrow g(t, x) = \mathbb{E}^{\mathbb{P}^*} \left[e^{-\int_t^T r(s, X_s) ds} G(X_T) \mid X_t = x \right]$$

$$\begin{aligned} \bullet \quad dX_t &= (u^x(t, X_t) - \lambda(t, X_t) \sigma^x(t, X_t)) dt \\ &\quad + \sigma^x(t, X_t) dW_t^* \end{aligned}$$

\bullet W_t^* is a \mathbb{P}^* -B.mtn.

??
 \rightarrow $dW_t^* = \lambda(t, X_t) dt + dW_t$

$$dX_t = u^x(t, X_t) dt + \sigma(t, X_t) dW_t$$

← "IP"
 the original
 SDE for X!

Girsanov's Theorem:

$$\frac{d\mathbb{P}^*}{d\mathbb{P}} = \mathbb{E} \left(\int_0^T \lambda_u dW_u \right)$$

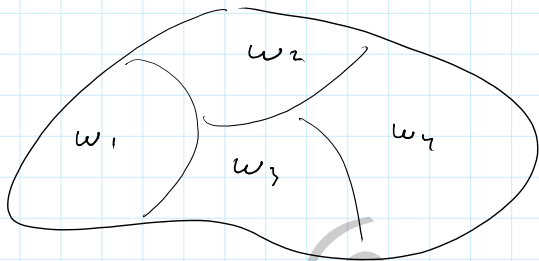
W_t is a \mathbb{P} -B.mtn.

↳ Doléan-Dade exponential

(stochastic exponential)

Then $W_t^* = -\int_0^t \lambda_u du + W_t$ is a \mathbb{P}^* -B.m.k.

$$\mathbb{P}: \Omega \mapsto [0, 1] \quad (\Omega, \mathcal{F}, \mathbb{P})$$



$$\begin{aligned} \mathbb{E}[Y] &= \int Y(\omega) d\mathbb{P}(\omega) \\ &= \sum Y(\omega_i) \mathbb{P}(\omega_i) \\ &= \sum \left(Y(\omega_i) \frac{\mathbb{P}(\omega_i)}{\mathbb{P}^*(\omega_i)} \right) \mathbb{P}^*(\omega_i) \\ &= \int Y(\omega) \left(\frac{d\mathbb{P}}{d\mathbb{P}^*} \right) (\omega) d\mathbb{P}^*(\omega) \end{aligned}$$

$\mathbb{P}(\omega_i) = 0$ if $\mathbb{P}^*(\omega_i) = 0$

$\mathbb{P}^*(\omega_i) = 0$ if $\mathbb{P}(\omega_i) = 0$ Radon-Nikodym derivative

$$\mathbb{E}^{\mathbb{P}^*}[Y] = \mathbb{E}^{\mathbb{P}} \left[Y \cdot \frac{d\mathbb{P}^*}{d\mathbb{P}} \right] = \mathbb{E}^{\mathbb{P}^*} \left[Y \cdot \frac{d\mathbb{P}}{d\mathbb{P}^*} \right]$$

e.g.:

$$Z \sim \mathbb{P} \sim \mathcal{N}(0, 1)$$

$$\eta \triangleq \frac{d\mathbb{P}^*}{d\mathbb{P}} = e^{-\frac{1}{2}\lambda^2 + \lambda Z}$$

clearly $\eta > 0$ a.s.

Radon-Nikodym derivative must

$$i) \mathbb{P} \left(\frac{d\mathbb{P}^*}{d\mathbb{P}} > 0 \right) = 1$$

$$ii) \mathbb{E}^{\mathbb{P}} \left[\frac{d\mathbb{P}^*}{d\mathbb{P}} \right] = 1$$

$$\begin{aligned} \mathbb{E}^{\mathbb{P}}[\eta] &= e^{-\frac{1}{2}\lambda^2} \mathbb{E}^{\mathbb{P}}[e^{\lambda Z}] \\ &= e^{-\frac{1}{2}\lambda^2} \cdot e^{\frac{1}{2}\lambda^2} = 1 \end{aligned}$$

$$\begin{aligned} \mathbb{E}^{\mathbb{P}^*}[e^{uZ}] &= \mathbb{E}^{\mathbb{P}} \left[e^{uZ} \frac{d\mathbb{P}^*}{d\mathbb{P}} \right] \\ &= \mathbb{E}^{\mathbb{P}} \left[e^{uZ} \cdot e^{-\frac{1}{2}\lambda^2 + \lambda Z} \right] \\ &= \mathbb{E}^{\mathbb{P}} \left[e^{(u+\lambda)Z} \right] e^{-\frac{1}{2}\lambda^2} \\ &= e^{\frac{1}{2}(u+\lambda)^2} e^{-\frac{1}{2}\lambda^2} = e^{u\lambda + \frac{1}{2}u^2} \end{aligned}$$

$$\left(\dots + \frac{1}{2}u^2 + u\lambda + \frac{1}{2}\lambda^2 - \frac{1}{2}\lambda^2 \right) \rightarrow \mathbb{P}^* \sim \mathcal{N}(\lambda, 1)$$

$$= e \quad e^{-} = e$$

$$\left(\text{expect} = \frac{1}{2} u^2 + u\lambda + \frac{1}{2}\lambda^2 - \frac{1}{2}\lambda^2 \right) \quad Z \stackrel{\mathbb{P}^*}{\sim} \mathcal{N}(\lambda, 1)$$

$$\mathcal{N}(0,1) \stackrel{\mathbb{P}^*}{\sim} Z^* \triangleq Z - \lambda$$

$$\mathbb{E} \left(\int_0^t \lambda_u dW_u \right) \triangleq e^{-\frac{1}{2} \int_0^t \lambda_u^2 du} + \int_0^t \lambda_u dW_u$$

$$\left(e^{-\frac{1}{2} \lambda^2} + \lambda Z \right)$$

$\leftarrow d$
 $\lambda = \text{const},$
 $t = 1$

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recall that to locally remove risk:

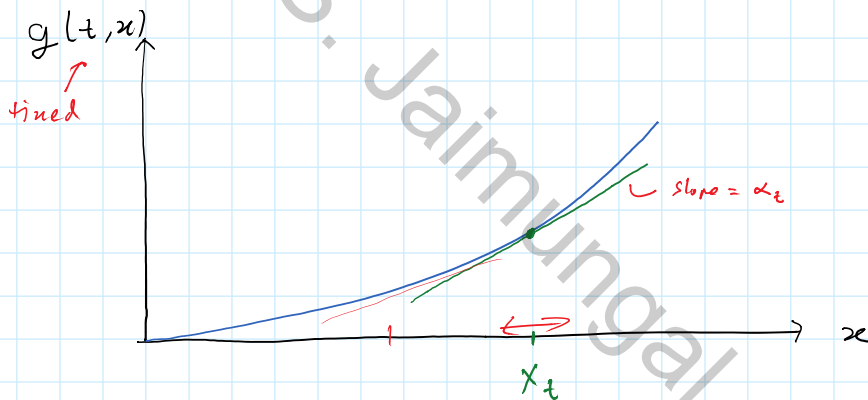
$$\alpha_t = \frac{g_t \sigma_t^g}{f_t \sigma_t^f}$$

if X_t is traded and so $f_t = X_t \Rightarrow$

$$\alpha_t = \frac{g_t \sigma_t^g}{\sigma_t^X}, \quad g_t \sigma_t^g = \partial_x g(t, X_t) \cdot \sigma_t^X$$

from Ito's lemma

$$\Rightarrow \alpha_t = \partial_x g(t, X_t)$$



how to use the "alpha" to hedge option via trading?

1) sell option gets $g_0 = g(0, X_0)$

buy α_0 of X , cost $\alpha_0 X_0$

bank account we have $B_0 = g_0 - \alpha_0 X_0$

at t_1 : α_0 of X now worth $\alpha_0 X_{t_1}$

bank account " has $B_{t_1} = (g_0 - \alpha_0 X_0) e^{r \Delta t}$

rebalance $\alpha_0 \rightarrow \alpha_{t_1}$ of X

profit/loss = $(\alpha_0 - \alpha_{t_1}) X_{t_1}$

$B_1 = (g_0 - \alpha_0 X_0) e^{r \Delta t} + (\alpha_0 - \alpha_{t_1}) X_{t_1}$

$$\text{profit/loss} = (\alpha_0 - \alpha_{t_1}) X_{t_1}$$

$$B_{t_1} = (q_0 - \alpha_0 X_0) e^{r\Delta t} + (\alpha_0 - \alpha_{t_1}) X_{t_1}$$

at t_2 :

α_{t_1} of X now with $\alpha_{t_1} X_{t_2}$

$$B_{t_2} = B_{t_1} e^{r\Delta t}$$

$\alpha_{t_1} \rightarrow \alpha_{t_2}$ of X , $\text{PnL} = (\alpha_{t_1} - \alpha_{t_2}) X_{t_2}$

$$B_{t_2} = B_{t_1} e^{r\Delta t} + (\alpha_{t_1} - \alpha_{t_2}) X_{t_2}$$

at t_k :

$\alpha_{t_{k-1}} \rightarrow \alpha_{t_k}$ of X ,

$$B_{t_k} = B_{t_{k-1}} e^{r\Delta t} + (\alpha_{t_{k-1}} - \alpha_{t_k}) X_{t_k}$$

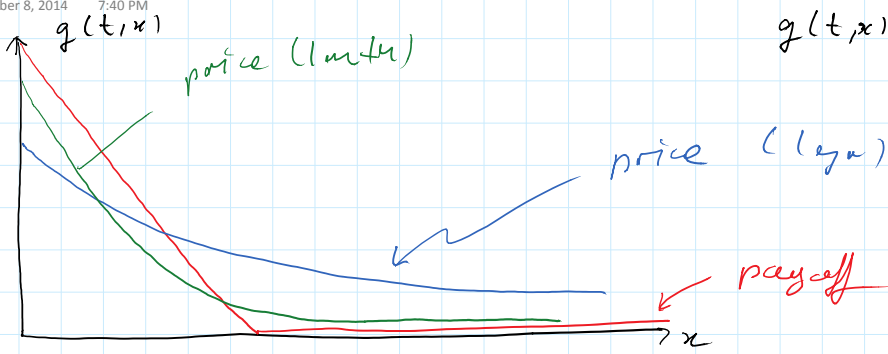
at $t_n = T$:

$\alpha_{t_{n-1}}$ of X

$B_{t_{n-1}} e^{r\Delta t}$ in Bank

owe $G(X_{t_n})$

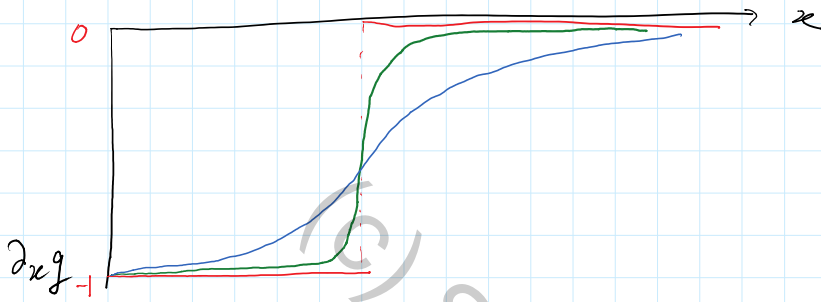
$$\text{PnL} = (\alpha_{t_{n-1}} X_{t_n} + B_{t_{n-1}} e^{r\Delta t}) - G(X_{t_n})$$



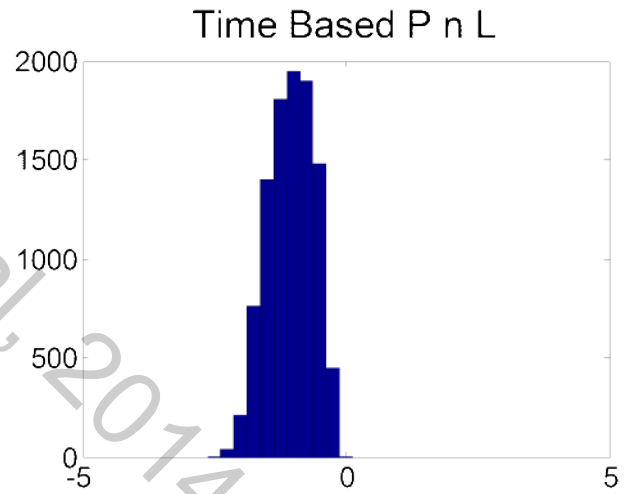
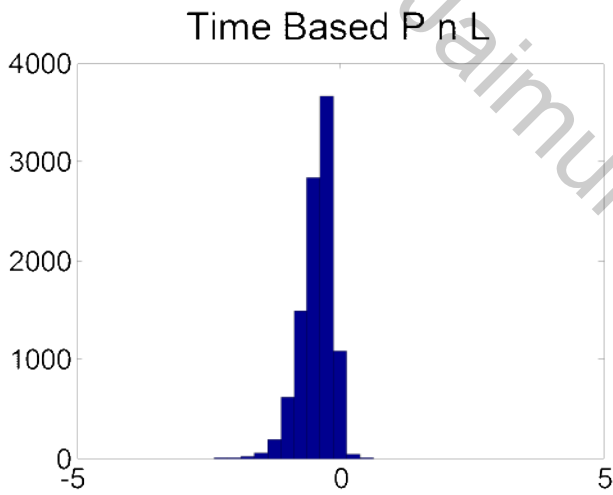
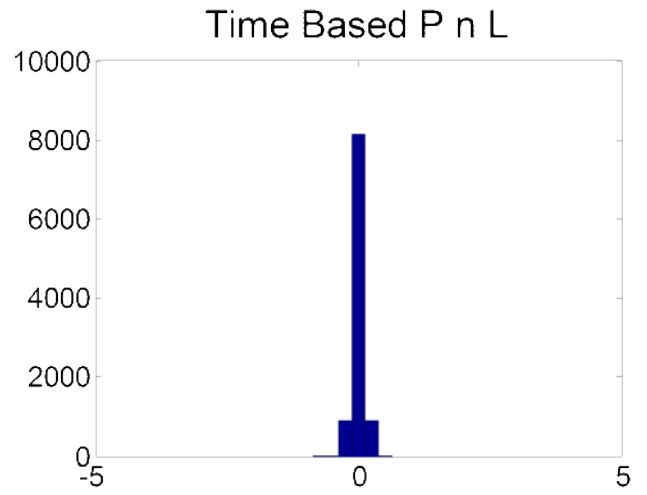
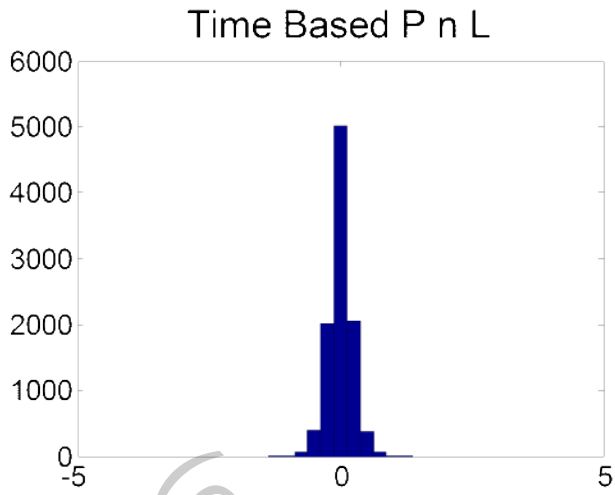
$$g(t, x) = K e^{-r\tau} \Phi(-d_+) - x \Phi(-d_-)$$

$$d_{\pm} = \frac{\log(x/K) + (r \pm \frac{1}{2}\sigma^2)\tau}{\sigma\sqrt{\tau}}$$

$$\tau = T - t$$



$$\partial_x g(t, x) = \Phi(-d_-) - 1$$



The "Greeks"

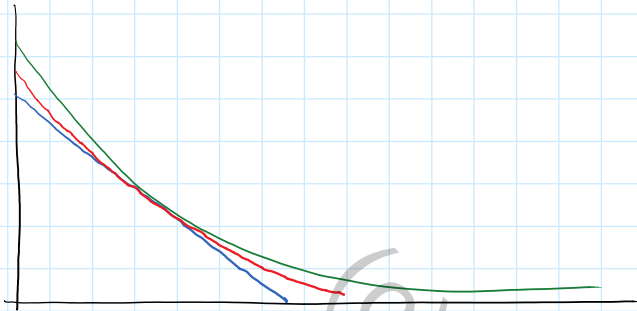
Δ_t^g - Delta

$$g(t, X_t + \Delta X_t) = g(t, X_t) + \Delta X_t \frac{\partial g}{\partial x}(t, X_t) + \frac{1}{2} (\Delta X_t)^2 \frac{\partial^2 g}{\partial x^2}(t, X_t) + \dots$$

Γ_t^g - Gamma

hedging option

$$V_t^h = \alpha_t X_t + \beta_t B_t + \eta_t h(t, X_t)$$



$$\begin{aligned} \frac{\partial V_t^h}{\partial x} &= \alpha_t + \eta_t \frac{\partial h}{\partial x} = \alpha_t + \eta_t \Delta_t^h = \Delta_t^g \\ \frac{\partial^2 V_t^h}{\partial x^2} &= \eta_t \frac{\partial^2 h}{\partial x^2}(t, X_t) = \eta_t \Gamma_t^h = \Gamma_t^g \end{aligned}$$

$$\eta_t = \frac{\Gamma_t^g}{\Gamma_t^h} \quad \alpha_t = \Delta_t^g - \frac{\Gamma_t^g}{\Gamma_t^h} \Delta_t^h$$

For us to be Delta-Gamma neutral.

$$V_t = \alpha_t X_t + \beta_t B_t + \eta_t h_t - g_t$$

$$\frac{\partial V_t}{\partial x} = 0$$

$$\frac{\partial^2 V_t}{\partial x^2} = 0$$