



Введение

Лекция Альбертовых мастеров

Стochastic mathematical mathematics:

стат. модели с. разн. \rightarrow УЧП \rightarrow ЧМ \rightarrow Проф.

+ экон. и фин. понимание

Asset Classes (of Tradable Assets)

(класс активов)

штучно продать

1 Equity

(акции)

Более рисковано

public
Equity
F GM
 ~ 50 трилл.

private

Equity

Более
стабильные
технологич.
рентаб.

"активы"
коэффициент
безопасности
богатства не подвергн.
(ML инвест. фонды)

2

Interest Rate Products

IRP
FI
Fixed Income

штучн. прод. ставок

Bonds

(корпор., местн., ...)

Chapter 11 - штучные формы банкр. в Америке,
капиталные модели функции давления

Более
предсказуемо

LIBOR Futures

(бумаги)

SWAPS

BIS
 ~ 200 трилл.

фин. контракт между участниками

длительный, один изобр. платит фикс. %
платежи, другой - плав. ставка

Прод. ставки - не фин. инструмент,

(экв. динамирование
отмуда берутся?)

Денеги - платежи

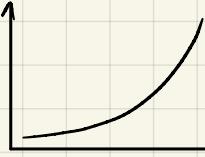
стоимость
капитала

$X(t) \sim x(t)$

$\dot{X}(t) = \alpha X(t) \Rightarrow \alpha = \text{const}, X = X_0 e^{\alpha t}$

$\Rightarrow \alpha = d(t), X = X_0 e^{d(t) \cdot t}$

$X(t) -$ все реал. (и не реал.) активы.



\Rightarrow долгосрочный и растущий долгущий ставки

\Rightarrow новая прод. ставка, т.е. долгущий эксп. эксп. роста

$d = 0,04 \Rightarrow$ прод. ставка долгосрочная \sim 4% год.

На SWAPs есть очень большое кредит. риски.

$10^{-2} = 1\text{pp}$ percent point

$10^{-4} = 1\text{bp}$ basis point

3 Currencies (Fx)

XAU XPT

XAG XPD

USD CHF

RUB JPY

GBP EUR

4 Commodities

глоб. индексы

(2)-(4) - FICC

5 Credit Derivatives (кредитный риск)

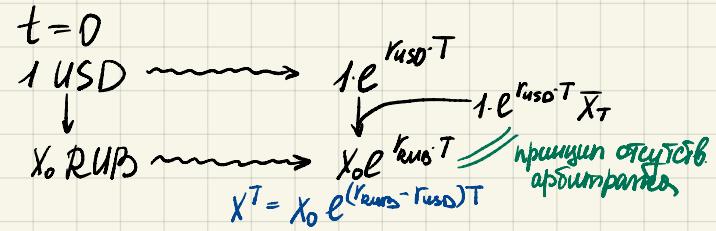
CDO

6* Digital Assets

$\text{BTC} \sim \text{XBT}$

USD/RUB
A B

USD: $r_{\text{USD}} = 0.25\%$
 $r_{\text{RUB}} = 4.25\%$
 $X_t = \frac{(\text{USD}/\text{RUB})_t}{x_t} \approx 75$



(1.1) - (1.2)

Ω - мн-бо беч норма-б
 $\Sigma \subseteq 2^\Omega$

- $\emptyset \in \Sigma, \Omega \in \Sigma$
- $A \in \Sigma \Rightarrow \bar{A} \in \Sigma$
- $\bigcup_{n=1}^{\infty} A_n \in \Sigma, \text{ even } \forall A_n \in \Sigma$

$P(A) \in [0,1]$ - мета на Σ , $P(\Omega) = 1$

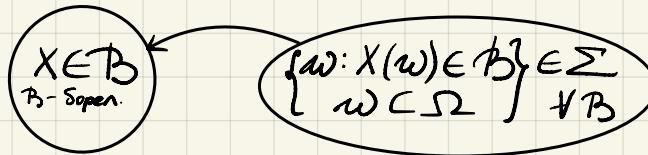
$$P\{U_{n=1}^{\infty} A_n\} = \sum_{n=1}^{\infty} P(A_n),$$

$A_i \cap A_j = \emptyset, i \neq j$

$$P(\emptyset) = X \Rightarrow P\{U_{n=1}^{\infty} \emptyset\} = \sum_{n=1}^{\infty} X, 0 \leq X \leq 1$$

$\Rightarrow X = 0$

$X: \Omega \rightarrow \mathbb{R}$

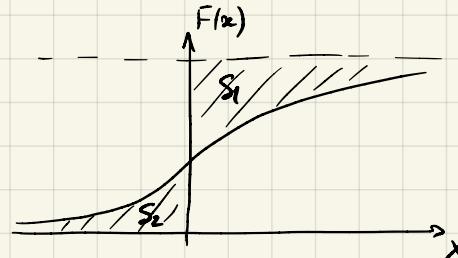


$S < t \Rightarrow \Sigma_s \subset \Sigma_t$

\Leftarrow - амн. бечено

$$\mathbb{E}X = \int_{\Omega} X(\omega) dP(\omega)$$

$$F(x) = \text{Prob}\{X \leq x\}$$



$$\begin{aligned} \mathbb{E}[x] &= S_1 - S_2 = \\ &= \int_0^{+\infty} (1 - F(x)) dx - \int_{-\infty}^0 F(x) dx \end{aligned}$$

$A \in \Sigma$
 $P(A) \rightsquigarrow \tilde{P}(A)$

$Z(\omega) \geq 0$
 $\mathbb{E}Z = 1 \Rightarrow \int_A Z(\omega) dP(\omega) \geq 0 \text{ almost surely}$

но в. Попов - Киприани
 эта ф-я не является изв. функц. и неодн.

$X \geq 0 \text{ almost surely}$

$\mathbb{E}[X] = \mathbb{E}[ZX]$

$Y \geq 0 \quad \mathbb{E}[Y] = \tilde{E}\left[\frac{Y}{Z}\right]$

$X \sim N(0,1) \quad f(x) = dF(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}$

$Y = X + \theta \sim N(\theta, 1)$

$Z(\omega) = e^{-X(\omega) \cdot \theta - \frac{\theta^2}{2}}$

$E[Z] = \int_{\Omega} Z(\omega) dF(\omega) =$

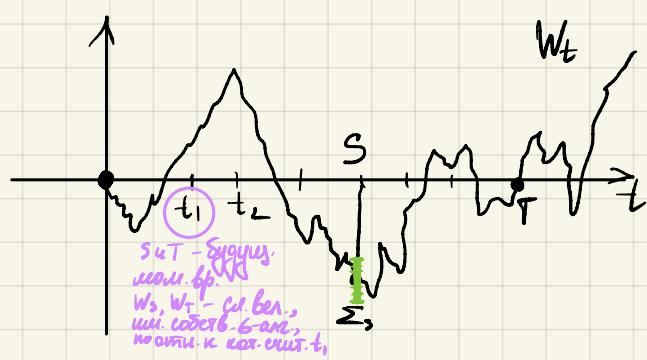
$$\int_{-\infty}^{+\infty} e^{-\theta x - \frac{\theta^2}{2}} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx = \frac{1}{\sqrt{\pi}} \int_{-\infty}^{+\infty} e^{-\frac{1}{2}(x^2 + 2\theta x + \theta^2)} dx = 1$$

$$\tilde{E}[Y] = E[ZY] = \int_{-\infty}^{+\infty} (x + \theta) e^{-\theta x - \frac{\theta^2}{2}} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx = \frac{1}{\sqrt{\pi}} \int_{-\infty}^{+\infty} (x + \theta) e^{-\frac{(x-\theta)^2}{2}} dx = 0$$

$$X \quad \Sigma \quad Y = X|\Sigma$$

$\forall A \in \Sigma$ - измеримая measurable
 $\int_A Y dP = \int_A X dP$
 условие н.о. для X

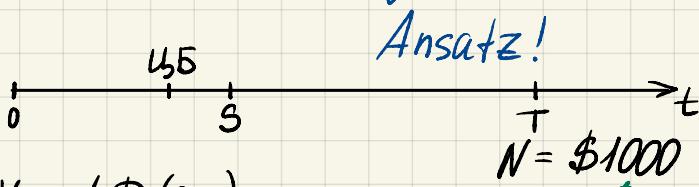
Ун. н.о. - использование
пространственных
функционалов.



$$\begin{aligned} \forall t \quad \mathbb{E}[W_t] &= 0 = \mathbb{E}[W_t | \Sigma_0] \\ t > s \quad \mathbb{E}[W_t | \Sigma_s] &= W_s - \text{нечёт.} - \text{ст. бег.} \\ T > s \quad \mathbb{E}[W_T | \Sigma_T] &= W_T \end{aligned}$$

свойство чётности
Броуновского движения

Обычное н.о. - фикс. величина
Численное н.о. - случай. величина



$$V_0 = N \cdot D(0, T)$$

н.о. дисонтирование

$$D(0, T) = \mathbb{E} \left[e^{-\int_0^T r_t dt} \right]$$

н.о. r_t -
постоянное уменьшающее
недифференциал.

Посчитать present value N

„Переоцененный годар покупать нечно,
потому надо его прибавить“

$$V_s = N \cdot D(s, T)$$

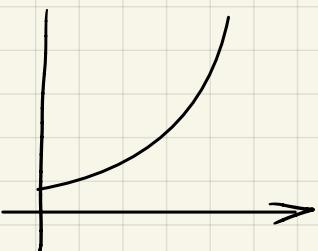
$$D(s, T) = \mathbb{E} \left[e^{-\int_s^T r_t dt} \right] = \mathbb{E} \left[e^{-\int_s^T r_t dt} | \Sigma_s \right] = f(r_s, s)$$

если прибавл., что для всех s ищ. макс,

$$\Sigma_s \subseteq \Sigma_t \subseteq \Sigma_T$$

$$\mathbb{E} \left[\mathbb{E} [X_T | \Sigma_t] | \Sigma_s \right] = \mathbb{E} [X_T | \Sigma_s]$$

Tower rule
(Rule of iterated condition)



$$x, \varphi(x)$$

$$\varphi(\mathbb{E}[x]) \leq \mathbb{E}[\varphi(x)]$$

Option

$$\frac{t=0}{S_0}$$

$$\frac{t=T}{S_T}$$

European Call Option
Underwriter Holder
Right $\leftarrow k$
Buy

$$S_T > K$$

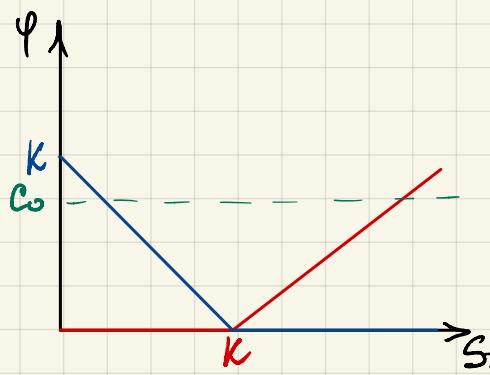
$$S_T - K$$

$$S_T \leq K$$

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

функция максимума

Contingent claim



$$\text{Eur. Call: } \Phi(S_T) = \max(S_T - K, 0) = (S_T - K)^+$$

$$\text{Eur. Put: } \Phi(S_T) = \max(K - S_T, 0) = (K - S_T)^+$$

$$C_0 = D(0, T) \mathbb{E}[\Phi(S_T)] + \xi_0 \geq \mathcal{A}(0, T) \varphi(\mathbb{E}[S_T]) + \xi_0$$

$$t_i = \tau \cdot i, i=0, 1, \dots$$

$$x_i = x(t_i)$$

$$x_0 = 0 \text{ c.b.p. 1}$$

$$x_{i+1} = x_i + \underbrace{\Delta_{i+1}}_{\sim N(0, \sqrt{\tau})}$$

$t_i \neq t_j$ Δ_i, Δ_j - independent

$$x_n = \sum_{i=1}^n \Delta_i$$

Random Walk
(eukrainejnye dyynegarue)

$$W_t = \lim_{\tau \rightarrow 0} x(t)$$

$$\mathbb{E}[X_n] = \sum \mathbb{E}[\Delta_i] = 0$$

$$\text{Var}[X_n] = \sum_{i=1}^n \underbrace{\text{Var}[\Delta_i]}_{\sigma^2} = n\sigma^2 = t_n$$

$$W_t \sim N(0, \sqrt{t})$$

$$\Delta W_{ij} = W_{tj} - W_{ti} \rightarrow \text{independent if } [t_i, t_j] \cap [t_k, t_l] = \emptyset$$

$$\Delta W_{kl} = W_{tl} - W_{tk}$$

$$3: W_s, t \geq s$$

$$\mathbb{E}[W_t | \Sigma_s] - ?$$

$$W_t = W_s + (W_t - W_s)$$

$$\mathbb{E}[W_t | \Sigma_s] = W_s + \mathbb{E}[W_t - W_s | \Sigma_s]$$

$$s < t$$

$$\text{Corr}(W_s, W_t) = \text{Corr}(W_s, W_s + \Delta W_{st}) = \frac{\text{Cov}(W_s, W_s + \Delta W_{st})}{\text{Std Dev}(W_s) \text{Std Dev}(W_s + \Delta W_{st})} =$$

$$= \frac{\mathbb{E}[W_s(W_s + \Delta W_{st})]}{\sqrt{st}} = \frac{\mathbb{E}[W_s^2 + W_s \Delta W_{st}]}{\sqrt{st}} = \frac{s + \mathbb{E}[W_s] \mathbb{E}[\Delta W_{st}]}{\sqrt{st}} = \sqrt{\frac{s}{t}} < 1$$

$$X_t = \mu t + \sigma W_t \sim N(\mu t, \sigma^2 t)$$

работает от 5 мин до часа, (short term)
головное приложение для генерации

USD/RUB Order Book

Asks (Orders)

10 5 3		\$5.02 ↑ mean
2 1		\$5.01 ↓
		Bid-Ask spread
3 5		\$4.99
6 7		\$4.98

$$VWAP (\text{Ask}, 10) = \frac{3}{10} \cdot \$5.01 + \frac{4}{10} \cdot \$5.02$$

$$VWAP (\text{Bid}, 10) = \frac{8}{10} \cdot \$4.99 + \frac{2}{10} \cdot \$4.98$$

Аспекто - участник, кот. хочет купить/продать
именно сейчас
(стартовый бирж.)

1.com = \$1000

(1.3) - (1.4)

$$X_t = \mu t + \sigma w_t$$

$$\begin{cases} \mu, \sigma = \text{const} \\ \sigma > 0 \end{cases}$$

Брауновское
движение

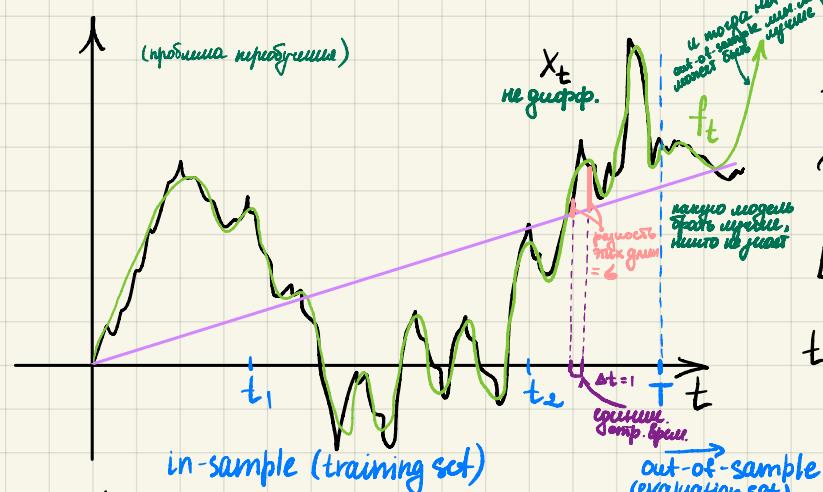
= 0 - не drift., < 0 - мед. мед., > 0 - мед. быст.

μ : Trend / Drift

σ : Volatility

"All models are wrong, but some models are useful"

w_t : Brownian motion $\Rightarrow -w_t$: also (by symmetry)



f_t : кусочно-гладкое

$$R_t = X_t - f_t$$

$$\mathbb{E} \left[\int_{t_1}^{t_2} R_t dt \right] = 0$$

$t_1 < t_2$ - random

распределение избога
нуль бюджета на предсказания

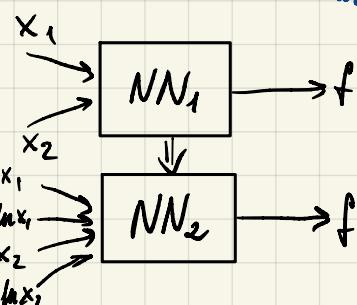
Фондоматематический и ма
математический "мир" (ML)

$$f_t = \mu t \Rightarrow X_t - \mu t = \sigma w_t$$

$$\mathbb{E} \left[\sigma \int_{t_1}^{t_2} w_t dt \right] = \sigma \lim_{\tau \rightarrow 0} \mathbb{E} \left[\sum_{n=1}^{\frac{t_2-t_1}{\tau}} W_n \cdot \tau \right] = 0$$

true price open,
 w_t - white

"мир" машинного обучения
стochastic. модели



$$f(x_1, x_2) - \infty - \text{diff.}$$

$$f = \frac{x_1}{x_2} !$$

$$dX_t = X_{t+dt} - X_t$$

$$dt \rightarrow 0$$

$$dW_t = W_{t+dt} - W_t \sim N(0, \sqrt{dt})$$

$$\boxed{dX_t = \frac{\mu dt}{\sigma(dt)} + \frac{\delta dW_t}{\sigma(dt)}}$$

$$\mathbb{E}[|dW_t|] = \sqrt{\frac{2}{\pi}} \sqrt{dt}$$

$$\begin{aligned} \mathbb{E}[|x|] &= \int_{-\infty}^{+\infty} |x| \varphi(x) dx = \int_{-\infty}^0 (-x) \varphi(x) dx + \int_0^{+\infty} x \varphi(x) dx = dy = -x \\ x \sim N(0, 1), \varphi(x) &= \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} = \int_{-\infty}^0 y \varphi(-y) f(dy) + \int_0^{+\infty} x \varphi(x) dx = \\ -2 \int_0^{+\infty} x \varphi(x) dx &= \sqrt{\frac{2}{\pi}} \int_0^{+\infty} x e^{-\frac{x^2}{2}} dx = \sqrt{\frac{2}{\pi}} = 0.8. \end{aligned}$$

$$\underline{dX_t \approx \delta dW_t}$$

$$|dX_t| = |\mu| dt + \delta |dW_t|$$

$$\mathbb{E}[|dX_t|] \leq |\mu| dt + \delta \sqrt{\frac{2}{\pi} dt} = Q(dt^{\frac{1}{2}})$$

Geometric Brownian Motion

GBM:

$$dX_t = \frac{\tilde{\mu}}{\tilde{\sigma}^2} X_t dt + \frac{\tilde{\sigma}}{X_t} dW_t$$

$$X_t = \text{USD/RUB}$$

$$\begin{bmatrix} X_0 \approx 45 \\ \tilde{\sigma} = 0.15 \end{bmatrix} \delta = 11.25$$

$$\tilde{\mu} = 0.0425 - 0.0025 = 0.04 \quad] \mu = 3$$

$$\text{Однако } \tilde{\sigma} > \tilde{\mu}$$

$$T = 1 \text{ h (hour)}$$

$$X_i, i = \frac{0, N}{\tau}$$

$$t: 0 \dots \frac{T}{\tau} \dots T$$

X_t - не единич. процесс

$$\Delta X_t: \Delta X_i = X_i - X_{i-1}, i = 1, \dots, N$$

$$\Delta X_i = \frac{\mu(t_i - t_{i-1})}{\tau} + \delta \Delta W_i$$

$$\Delta X_i \sim N(\mu\tau, \delta^2\tau)$$

$$\Delta X_i: \text{H.D.P.C.B.}$$

$$\begin{aligned} \hat{\mu} &= \frac{1}{\tau} \mathbb{E}[\Delta X] = \frac{1}{N\tau} \sum_{i=1}^N \Delta X_i = \frac{1}{T} \sum_{i=1}^N \Delta X_i = \\ &= \frac{1}{T} (X_T - X_0) \end{aligned}$$

$$X_0 = 0, \quad \hat{\mu} = \frac{X_T}{T} - \text{искусст. оценка}$$

$$\begin{aligned} \mathbb{E}[\hat{\mu}] &= \frac{M}{\tau} \\ \text{Var}[\hat{\mu}] &= \frac{M}{\tau^2} \mathbb{E}[X_T^2] - (\mathbb{E}[\hat{\mu}])^2 = \\ &= \mathbb{E}[\hat{\mu}^2] - \mu^2 = \frac{M}{\tau^2} \mathbb{E}[X_T^2] - M^2 \quad \text{---} \\ X_T^2 &= \frac{M^2 T^2}{\tau^2} - \frac{2M\delta TW_T + \delta^2 W_T^2}{\tau^2} \\ \mathbb{E}[X_T^2] &= \frac{M^2 T^2}{\tau^2} + 0 + \delta^2 T \\ &\Rightarrow \frac{\mu^2 T^2 + \delta^2 T}{\tau^2} - \mu^2 = \frac{\delta^2}{T} \neq f(N)! \end{aligned}$$

$$\Delta X \sim N(M\bar{t}, \sigma^2)$$

$$\frac{x_t}{N} = \bar{x}$$

$$\sum_{i=1}^N (\Delta x_i - \frac{x_t}{N})^2 - \text{бесхиле кесимүз айынка}$$

$$\mathbb{E}[\hat{\sigma}^2] = \sigma^2$$

$$\text{Var}[\hat{\sigma}^2] = \text{Var}(\frac{1}{N} \sum_{i=1}^N (\Delta x_i - \frac{x_t}{N})^2) \sim \text{large Var}$$

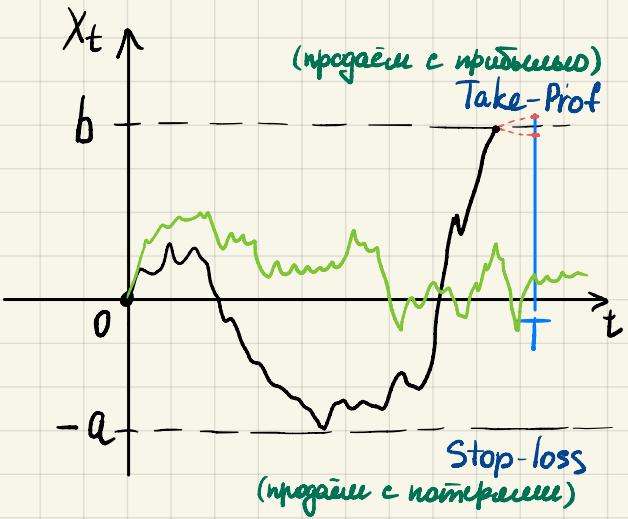
$$\hat{\sigma}^2 = \frac{1}{N^2} \sum_{i=1}^N \Delta x_i^2 - \text{сипел. айынка, асимилят. шиеру} \rightarrow 0$$

Realized Variance
smaller Var

$$\mu = 0 \quad X_t = 6W_t \quad \text{мартиник}$$

$$\mathbb{E}[X_t | \sum_s] = X_s = 6W_s$$

$$\mathbb{E}[X_t] = \mathbb{E}[X_t | \sum_t] = 0$$



First time:

Event t^* : $X_{t^*} = b$ or $X_{t^*} = -a$
 $\exists t^*$ almost surely

$$P = P[X_{t^*} = b] \Rightarrow 1 - P = P[X_{t^*} = -a]$$

$$0 = \mathbb{E}[X_{t^*}] = \text{at } t^* = pb - a(1-p) = p(a+b) - a \Rightarrow \text{безрисковая стратегия}$$

$$P = \frac{a}{a+b}$$

$$B: \exists 0 < t^* \leq T : X_{t^*} = b$$

кеналуулук принцип симметрии
(метод зеркальности)

$$\Phi(x) = \int_{-\infty}^x \frac{1}{\sqrt{2\pi}} e^{-\frac{u^2}{2}} du. \quad X_t = 6W_t \sim N(0, 6\sqrt{t})$$

$$P[X_T \leq x] = P\left[\frac{X_T}{6\sqrt{T}} \leq \frac{x}{6\sqrt{T}}\right] = \Phi\left(\frac{x}{6\sqrt{T}}\right)$$

$$P[X_T \leq b] = \Phi\left(\frac{b}{6\sqrt{T}}\right) = P[X_T \leq b | B]P + P[X_T \leq b | \bar{B}](1-P) = \frac{1}{2}P + 1 - P = 1 - \frac{P}{2}$$

$$P = 2 \left[1 - \Phi\left(\frac{b}{6\sqrt{T}}\right) \right]$$

$$\begin{cases} b, P = P \\ X_T, P = 1 - P \end{cases}$$

$$\mathbb{E}[P_n L] = bP + (1-P)\mathbb{E}[X_T | B]$$

нодушлат, как наурует
 $B: \max_{0 \leq t \leq T} X_t < b$

Как нодушрат $b \in T$, шодын науке $P_n L$

$$dX_t = \mu dt + \sigma dW_t \Rightarrow X_t = \mu t + \sigma W_t$$

$$dX_t = \mu_t dt + \sigma_t dW_t \Rightarrow X_t = \int_0^t \mu_s ds + ?$$

$\int_0^t \sigma_s dW_s$

Y to Integral

$$\int_0^t \sigma_s dW_s = \lim_{\delta \rightarrow 0} \sum_{i=1}^N \sigma(t_{i-1})(W_{t_i} - W_{t_{i-1}}) = I_t$$

noxonee na unen. Cmienem.
no b. gaujome ukr. p-yue, no kov. ukr. ukr. ukr.

ne ukr. gaujome ukr.

Also: σ_t - Random measurable \sum_t
 dW_t independent of σ_t

$$\begin{aligned} \mathbb{E}[I_t] &= 0 \\ \text{Var}[I_t] &= \lim_{\delta \rightarrow 0} \left(\sum_{i=1}^N \mathbb{E}[\sigma_i^2 \Delta W_i^2] \right) + \sum_{i,j=1}^N \mathbb{E}[\sigma_i \sigma_j \Delta W_i \Delta W_j] = \lim_{\delta \rightarrow 0} \sum_{i=1}^N \mathbb{E}[\sigma_i^2] \delta t = \\ &\quad \mathbb{E}[\sigma_i^2] \delta t: \quad \begin{matrix} \sigma_i \Delta W_i \\ \text{indep.} \end{matrix} \quad \begin{matrix} \sigma_j \Delta W_j \\ \text{indep.} \end{matrix} \\ &\quad \mathbb{E}[\sigma_i \sigma_j \Delta W_i] \otimes (\Delta W_j) \end{aligned}$$

$$\sigma_t - \text{determ: } \mathbb{E}[I_t] = \lim_{\delta \rightarrow 0} \sum_{i=1}^N \sigma(t_i) \mathbb{E}[\Delta W_i] = 0$$

$$\begin{aligned} \text{Var}[I_t] &= \lim_{\delta \rightarrow 0} \sum_{i=1}^N \text{Var}[\sigma(t_i) \Delta W_i] = \lim_{\delta \rightarrow 0} \sum_{i=1}^N \sigma^2(t_i) \text{Var}[\Delta W_i] \\ &\quad \sigma^2(t_i) \text{Var}[\Delta W_i] \\ &\quad \sigma^2(t_i) (t_i - t_{i-1}) \end{aligned}$$

$$\int_0^t \sigma^2(s) ds$$

Numerical Nimo cibveemce markovianade

$$\mathbb{E}[I_t | \sum_s] = \mathbb{E}\left[\int_0^s \sigma_u dW_u | \sum_s\right] + \mathbb{E}\left[\int_s^t \sigma_u dW_u | \sum_s\right] = \underline{\int_0^s \sigma_u dW_u} = I_s$$

$$X_t = \mu_t dt + \sigma_t dW_t \Leftrightarrow X_t = \int_0^t \mu_s ds + \int_0^t \sigma_s dW_s$$

$$\begin{cases} \mu_t = \mu(X_t, t) \\ Y_t = f(X_t, t) \end{cases} \sum_t$$

\hat{Y}_t - Döblin
1942 1940

$$dY_t = \frac{\partial f}{\partial X} dX_t + \frac{\partial f}{\partial t} dt + \frac{1}{2} \frac{\partial^2 f}{\partial X^2} (dX_t)^2$$

$$(dX_t)^2 = \mu_t^2 dt^2 + 2\mu_t \sigma_t dt dW_t + \sigma_t^2 dW_t^2$$

$$O(dt^{1/2}) \rightarrow 0$$

$$\mathbb{E}[dW_t^2] = dt$$

$$\sigma_t^2 dt$$

$$\mathbb{E}\left[\int_0^T dW_s^2\right] = ? = T$$

$$\lim_{\tau \rightarrow 0} \sum_{i=1}^N (W_{t_i} - W_{t_{i-1}})^2$$

$$t_i = \tau \cdot i, T = \tau \cdot N$$

$$\text{Var}[dW_t^2] = \mathbb{E}[dW_t^4] - \mathbb{E}[dW_t]^2 =$$

$$= 3dt^2 - dt^2 = 2dt^2$$

$$\text{Std Dev}[dW_t^2] = \sqrt{2} dt$$

$$\begin{aligned} \text{Var}\left[\int_0^T dW_s^2\right] &= \mathbb{E}\left[\left(\int_0^T dW_s^2\right)^2\right] - T^2 = \\ &= \lim_{\tau \rightarrow 0} \left(\sum_{i=1}^N \mathbb{E}[(dW_i^2)^2] + 2 \sum_{i=1}^N \sum_{j=i+1}^N \mathbb{E}[dW_i^2 \cdot dW_j^2] \right) - T^2 = \\ &= \lim_{\tau \rightarrow 0} \left(3\tau^2 N + 2\tau^2 \frac{N(N-1)}{2} \right) - T^2 = \lim_{\tau \rightarrow 0} (2\tau^2 (3N - N^2 - N)) - T^2 = \\ &= \lim_{\tau \rightarrow 0} (2N\tau^2 + N^2\tau^2) - T^2 = \lim_{\tau \rightarrow 0} (2\tau^2) = 0 \end{aligned}$$

Rewriting Rule: $dW_t^2 \rightarrow dt$

(1.5) - (1.7)

Dividends:

→ continuous

→ Relative to curr stock price

$$D_t S_t dt$$

dividend rate

если начали платить дивиденды,

то price становится выше и увеличивается

избрать со временем
новой структуры компании
(и уменьшить)

$$dS_t = (\mu_t - D_t) S_t dt + \frac{\sigma_t S_t dW_t}{\delta(S_t, t)}$$

FX: $P_x(A / B)$ = # of B units
Asset
equir. by value to 1 unit of A
Base Ccy / Quote Ccy
currency Settlement Ccy

EUR GBP USD
AUD CHF

Asset
AAPL / USD
Instrument

A / B

$$D_t = r_A$$

$$\mu_t = r_B$$

$$S_t = P_x(A/B)$$

$$dS_t = (r_B - r_A) S_t dt + \frac{\sigma_t S_t dW_t}{\text{or more compact}}$$

$$\begin{aligned} S_t &[\text{RUB}] \\ r_A, r_B &[\text{year}^{-1}] \\ dt &[\text{year}] \\ dW_t &[\text{year}^{1/2}] \\ \Rightarrow \sigma_t &[\text{year}^{-1/2}] \end{aligned}$$

как и в фьюнк.,
помогло проверить
длительности

$$\text{ACT/365} = \frac{\Delta t \text{ (days)}}{365 \text{ Fixed Len}}$$

$$\text{ACT/360} = \dots \quad (\text{где упрощ. - 12 мес. и 30 дней})$$

$$\text{ACT/365.25} = \dots \quad \text{Среднее по всем} - 365.2425$$

$$365.25 - \frac{3}{400}$$

$$\text{ACT/352} = \dots \quad \text{с учетом празднич. дня: } 365.2423..$$

$$\text{VOL} = 15\% / \text{year}^{1/2} = \frac{15\%}{(252 bd)^{1/2}} \approx \frac{1\%}{bd^{1/2}}$$

стажировки:

Quant/ Stock	Options Val.	QA	Quantitative Analyst
	Equity Portfolio	QR	Quantitative Researcher
	Data engineer	RA	Research Analyst

$$dS_t = \mu_t S_t dt + \sigma_t S_t dW_t$$

S_t : commodity price
 $S_t > 0$

$$dS_t = \alpha(\theta - S_t) dt + \sigma_t S_t dW_t$$

множ. бывш. времени

эта модель
аналит. решения не имеет.
Но явные решения не однозначны.
Гораздо более находит практика расчет S_t

$$dB_t = \alpha(\theta - S_t e^{-dt}) dt + \dots$$

Откуда брать α ? — ценовую? опционную? цену commodities

$$\vec{q} = \begin{bmatrix} \frac{20}{3} \\ \frac{1}{3} \\ \frac{1}{3} \\ \frac{1}{3} \end{bmatrix} \in \mathbb{R}^5$$

$$t_0 \dots t_N = T$$

$$T = 3 \text{ m}$$

$$288.65 = 18420 = N$$

$$T = 5 \text{ min}$$

$$\uparrow \text{ торговые дни}$$

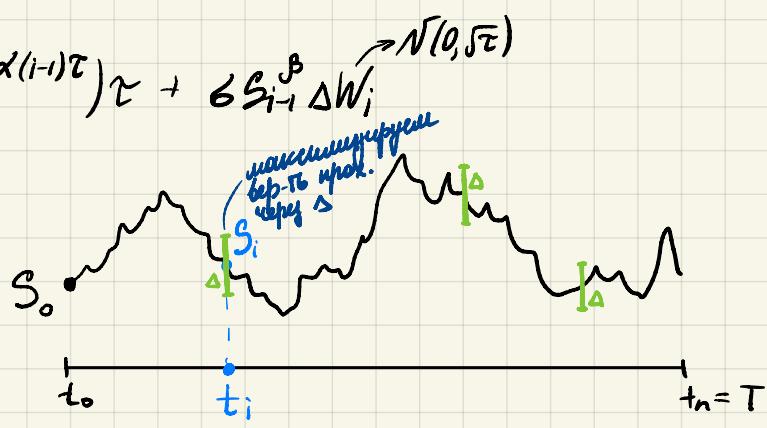
Метод гаусс. фильтр. гауссовое к-во.

$$dS_i \approx S_i - S_{i-1}; \quad S_i \approx S_{i-1} + \alpha(\theta - S_{i-1} e^{-\alpha(i-1)\tau}) + \sigma S_{i-1}^{\beta} \Delta W_i \sim N(0, \sigma^2)$$

Метод макс. правдоподобия

~~надежность~~

$$\begin{aligned} & p(S_1, t_1 | S_0) \Delta \\ & p(S_2, t_2 | S_0, S_1) \Delta \quad \text{Марковское прошлое} \\ & p(S_i, t_i | S_0, \dots, S_{i-1}) \Delta \quad \text{ни-ни не яв. от истории} \end{aligned}$$



$$S_{ti} \sim N(S_{i-1} + \alpha(\dots)\tau, \sigma^2 S_{i-1}^{\beta} \sqrt{\tau})$$

$$p(S_i, t_i | S_{i-1}) =$$

$$\begin{aligned} & p(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}. \quad F(S_i) = \Phi\left(\frac{S_i - M}{\sigma}\right) \\ & f(S_i) = F'(S_i) = \frac{1}{\sigma} \Phi\left(\frac{S_i - M}{\sigma}\right) \\ & = \varphi\left(\frac{S_i - S_{i-1} - \alpha(\dots)\tau}{\sigma S_{i-1}^{\beta} \sqrt{\tau}}\right) \frac{1}{\sigma S_{i-1}^{\beta} \sqrt{\tau}} \end{aligned}$$

$$\begin{aligned} L &= \prod_{i=1}^N (p(S_i, t_i | S_{i-1})) \Delta \\ \tilde{L} &= \ln L = N \ln \Delta = \sum_{i=1}^N \ln p(S_i, t_i | S_{i-1}) \end{aligned}$$

решение можно получить
таким же способом

$$\alpha > 0$$

Оптимизируем модель \rightarrow значение $f(\Delta W_i)$ \rightarrow стат. мерк на броуз. вв.

$$\sum_{i=k}^l \Delta W_i^2 \rightarrow t_l - t_k = \tau(l-k)$$

Запоминавши к-во в видах-ти даёт большую погр. ошибку, чтобы эти к-во не запоминать

$$X_t = f(\beta_t)$$

$$\frac{df}{dS} = \frac{1}{\sigma \beta_t^\beta} > 0$$

$$dS_t = \mu(\beta_t, t) dt + \sigma(S_t) dW_t$$

$$\begin{aligned} dX_t &= \frac{\partial f}{\partial \beta} dS_t + \frac{1}{2} \frac{\partial^2 f}{\partial S^2} \sigma^2 = \frac{1}{\sigma \beta_t^{\beta-1}} (\mu(S_t, t) dt + \sigma S_t^\beta dW_t) + \frac{1}{2} \frac{1}{\sigma} (-\beta) \frac{1}{\beta_t^{\beta-1}} \sigma^2 S_t^{\beta-1} dt = \\ &= \left(\frac{\mu(S_t, t)}{\sigma \beta_t^{\beta-1}} - \frac{\beta \sigma^2 S_t^{\beta-1}}{2} \right) dt + 1 \cdot dW_t \end{aligned}$$

$$\hat{\mu}(X_t) = (\dots) f^{-1}(X_t) \quad \xrightarrow{\text{замет. на 1-й счет усредн. тренда}} \text{метод гаусс. к-во}$$

$$\begin{aligned} \tilde{L} &= \sum_{i=1}^N \ln \hat{p}(X_i, t_i | X_{i-1}) \rightarrow \max \\ \hat{p}(X_i, t_i | X_{i-1}) &= \varphi(X_i - X_{i-1} - \hat{\mu}(X_{i-1}) \tau) \end{aligned}$$

Но есть ошибка!

$$\begin{aligned} dS_t &= \mu(S_t, t, \bar{q}) dt + \sigma(S_t, \bar{q}) dW_t \\ X_t &= f(S_t, \bar{q}): \frac{dt}{ds} = \sigma(S_t, \bar{q}) \end{aligned}$$

$$\begin{aligned} \tilde{\tau} &= \sum_{i=1}^N \ln \left(\frac{\hat{\mu}(X_i, t_i | X_{i-1})}{\sigma(S_i)} \right) \\ &\text{правильно!} \end{aligned}$$

$$\begin{array}{ccccccc} i & 0 \dots i \dots N & \leftarrow \\ S_i & S_0 \dots S_i \dots S_N & \\ X_i(\bar{q}) & & \end{array}$$

мы максимизуем другую ф-цию правдоподобия!

Помехи в новых
предсказаниях
не отменяют
усл. S_i

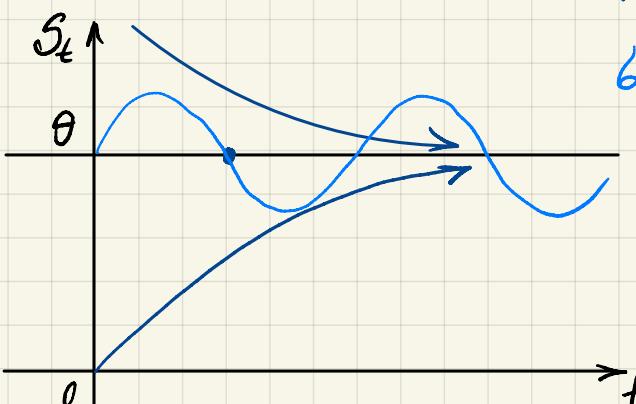
$$\mathbb{P}\{S_{t_i} \in (S_i - \Delta/2, S_i + \Delta/2)\} = \mathbb{P}\{f(S_{t_i}, \bar{q}) \in (f(S_i - \Delta/2, \bar{q}), f(S_i + \Delta/2, \bar{q}))\} =$$

$$= \mathbb{P}\{X_{t_i} \leq (f(S_i, \bar{q}) - f'(S_i, \bar{q})\frac{\Delta}{2}, f(S_i, \bar{q}) + f'(S_i, \bar{q})\frac{\Delta}{2})\} = p(X_i, t_i | X_{i-1}) \cdot \frac{1}{\delta(S_i, \bar{q})}$$

Непрерывный
интервал

$$A < 0 \quad \theta + A \sin(2\pi t + \varphi_0) \leftarrow \text{без эфекта синхронности}$$

$$dS_t = \alpha(\theta - S_t e^{-\alpha(t-t_0)}) dt + \sigma S_t^{\beta} dW_t$$



$$\theta \rightarrow \bar{S}_t = \mathbb{E}[S_t]$$

$$d\bar{S}_t = \alpha(\theta - \bar{S}_t e^{-\alpha(t-t_0)}) dt$$

Зависимость процесса S_t от стокастического интеграла dW_t

$$d\bar{S}_t = -\alpha e^{-\alpha(t-t_0)} \bar{S}_t dt \quad t \geq t_0$$

$$\frac{d\bar{S}_t}{\bar{S}_t} = -\alpha e^{-\alpha(t-t_0)} dt \quad T = t - t_0$$

$$\ln \bar{S}_T - \ln C = \frac{\alpha}{\alpha} (e^{-\alpha T} - 1) = \frac{\alpha}{\alpha} (1 - e^{-\alpha T})$$

$$\bar{S}_T = C e^{-\frac{\alpha}{\alpha} (1 - e^{-\alpha T})} \quad C = C(T) \Rightarrow \bar{S}(T) \quad \text{затухание}$$

$$X_t = A \sin \omega t + \sigma W_t$$



Ornstein - Uhlenbeck

$$dS_t = \alpha(\theta - S_t) dt + \sigma dW_t$$

$$\frac{dS_t}{dS_0} = \frac{p(S_t | S_0)}{S_t} \quad S_t = X_t e^{-\alpha t} \quad X_t = S_t e^{\alpha t} = f(S_t, t)$$

$$dX_t = S_t e^{\alpha t} \alpha e^{\alpha t} dt + e^{\alpha t} dS_t = e^{\alpha t} (S_t \alpha dt + \alpha(\theta - S_t) dt + \sigma dW_t) = e^{\alpha t} (\alpha \theta dt + \sigma dW_t)$$

$$X_t = X_0 + \alpha \theta \frac{1}{2\alpha} (e^{2\alpha t} - 1) + 6 \int_0^t e^{2\alpha s} dW_s$$

$$S_t = S_0 e^{-\alpha t} + \theta (1 - e^{-\alpha t}) + 6 \int_0^t e^{\alpha(t-s)} dW_s$$

$$\mathbb{E}[S_t] = S_0 e^{-\alpha t} + \theta (1 - e^{-\alpha t}) \quad \text{unif. boro} \sim N(0, 6 \sqrt{\frac{1-e^{-2\alpha t}}{2\alpha}})$$

$$\begin{aligned} \text{Var} &= 6^2 \int_0^t e^{-2\alpha(t-s)} ds \\ &= \left(\frac{6^2}{-2\alpha}\right) e^{-2\alpha(t-s)} / t \\ &= \frac{6^2}{2\alpha} (1 - e^{-2\alpha t}) \end{aligned}$$

$$\theta = \lim_{t \rightarrow \infty} \mathbb{E}[S_t]$$

$$\frac{6^2}{2\alpha} = \lim_{t \rightarrow \infty} \text{Var}[S_t]$$

$$S_t \sim N(S_0 e^{-\alpha t} + \theta (1 - e^{-\alpha t}), 6 \sqrt{\frac{1-e^{-2\alpha t}}{2\alpha}})$$

$$S_t | S_{i-1} \sim N(S_{i-1} e^{-\alpha(t-t_{i-1})} + \theta (1 - e^{-\alpha(t-t_{i-1})}), 6 \sqrt{\frac{1-e^{-2\alpha(t-t_{i-1})}}{2\alpha}})$$

$$p(S_i | S_{i+1}) = \varphi \left(\frac{S_i - S_{i-1} e^{-\alpha t} - \theta (1 - e^{-\alpha t})}{6} \right)$$

$$\bar{\sigma} = 6 \sqrt{\frac{1-e^{-2\alpha t}}{2\alpha}}, \quad \tau = t_i - t_{i-1}$$

$$\mathcal{L} = \sum_{i=1}^N \ln p(S_i | S_{i-1}) \longrightarrow \max_{\bar{q}}, \quad \bar{q} = \begin{bmatrix} \alpha \\ \theta \\ \bar{\sigma} \end{bmatrix}$$

$$\varphi(x) = \text{const.} \cdot e^{-\frac{x^2}{2}}$$

$$-\frac{1}{2} \sum_{i=1}^N \left(\frac{S_i - S_{i-1} e^{-\alpha t} - \theta (1 - e^{-\alpha t})}{6} \right)^2 - N \ln \bar{\sigma} \longrightarrow \max_{\bar{q}}$$

$$N \ln \bar{\sigma} + \frac{1}{2\bar{\sigma}^2} \sum_{i=1}^N (S_i - S_{i-1} e^{-\alpha t} - \theta (1 - e^{-\alpha t}))^2 \longrightarrow \min_{\bar{q}}$$

$$\begin{aligned} \mathbb{E}[\hat{x}] &\neq x & \mathbb{E}[\hat{x}] &\xrightarrow{N \rightarrow \infty} x && -\text{currens. u ac. currens.} \\ \mathbb{E}[\hat{\theta}] &= \theta & \text{Var}[\hat{\theta}] &\xrightarrow{\text{large, slowly}} 0 && \xrightarrow{N \rightarrow \infty} 0 \\ \mathbb{E}[\hat{\sigma}^2] &= \sigma^2 & \text{Var}[\hat{\sigma}^2] &\xrightarrow{\text{small}} 0 && \xrightarrow{N \rightarrow \infty} 0 \end{aligned}$$

Model: $\bar{q} \in \mathbb{R}^k$

(1) Generate $\{\bar{q}_j\}$
 (2) For each $\bar{q}_j \rightarrow$ Using Monte-Carlo
 generate $\{S_1^{(m)}, S_2^{(m)}\}_{M \text{ times}}$

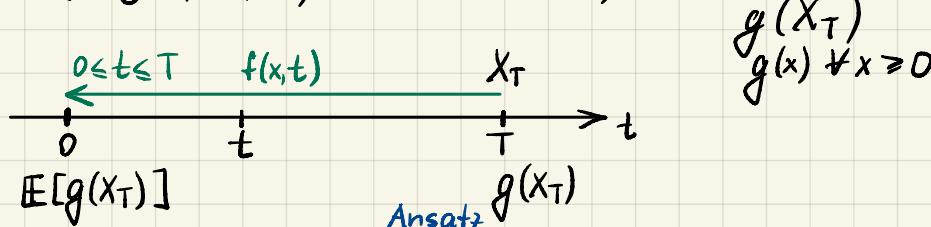
(3) Using MLE estimator estimate param.

(4) Get the means $\bar{q}_j = \sum_{m=1}^M \hat{q}_j^{(m)}$ $M \text{ times}$

(5) Construct the inverse: $\bar{q}_j = \frac{1}{\sum_{m=1}^M \hat{q}_j^{(m)}} \hat{q}_j^{(m)} \xrightarrow{\text{Neural Net}} \bar{q}_j$ no error term

Feynman-Kac PDE

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



$$\mathbb{E}[g(X_T) | \Sigma_t] = f(X_t, t)$$

$$\mathbb{E}[f(X_T, t) | \Sigma_s] = \mathbb{E}[\mathbb{E}[g(X_T) | \Sigma_t] | \Sigma_s] = \mathbb{E}[g(X_T) | \Sigma_s] = f(X_s, s)$$

$$df = \frac{\partial f}{\partial t} dt + \frac{\partial f}{\partial x} dX_t + \frac{1}{2} \frac{\partial^2 f}{\partial x^2} \sigma^2 dt = \left(\frac{\partial f}{\partial t} + \mu \frac{\partial f}{\partial x} + \frac{\sigma^2}{2} \frac{\partial^2 f}{\partial x^2} \right) dt + \frac{\partial f}{\partial x} \sigma dW_t$$

$f(x, t) - ?$

$$\Rightarrow \frac{\partial f}{\partial t} + \mu(x, t) \frac{\partial f}{\partial x} + \frac{\sigma^2(x, t)}{2} \frac{\partial^2 f}{\partial x^2} = 0$$

$\left[\begin{array}{l} f(x, T) = g(x), 0 \leq t \leq T \\ \text{convective term} \\ \text{diffusion term} \\ > 0 \end{array} \right]$

(== обратное уравнение Копенгорода)

$$f(0, t) = g(0) \quad \leftarrow \text{Dirichlet}$$

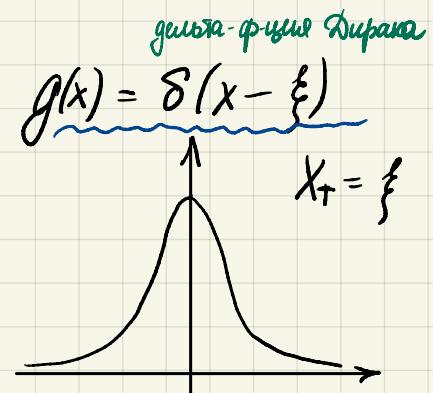
$$\lim_{x \rightarrow \infty} f'_x(x, t) = \lim_{x \rightarrow \infty} g'(x) \quad \leftarrow \text{Neumann}$$

$$\frac{\partial f}{\partial t} = -L_2 f \quad L_2 = \mu \frac{\partial}{\partial x} + \frac{\sigma^2}{2} \frac{\partial^2}{\partial x^2}$$

$$f(x, t) = \frac{1}{\sqrt{2\pi(T-t)\sigma^2}} e^{-\frac{(x-\xi)^2}{2(T-t)\sigma^2}}$$

$$\mu(x, t) = 0$$

$$\sigma(x, t) = \sigma = \text{const}$$



Fokker-Planck

$$\begin{cases} dX_t = \mu(X_t, t) dt + \sigma(X_t, t) dW_t \\ X(0) = X_0 \text{ a.s.} \end{cases}$$

$p(x, t)$: PDF X_t

$$h(X_T) = ? = \int_0^T dh(X_t) + h(X_0)$$

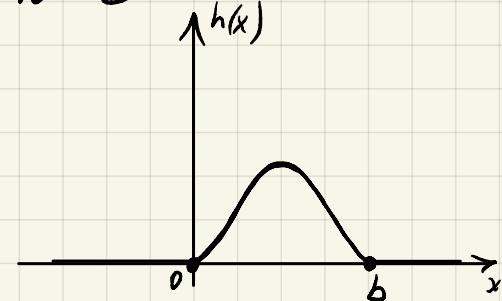
$$dh(X_T) = h'(X_t) dX_t + h''(X_t)/2 (dX_t)^2 =$$

$$= h'(X_t) \mu(X_t, t) dt + \frac{h''(X_t)}{2} \sigma^2(X_t, t) dt +$$

$$+ h'(X_t) \sigma(X_t, t) dW_t$$

$$\mathbb{E}[dh(X_t) | \Sigma_t]$$

$$\begin{aligned} h(x \leq 0) &= 0 \\ h(x \geq b) &= 0 \\ h(0 < x < b) &> 0 \\ h &\in C^2 \end{aligned}$$



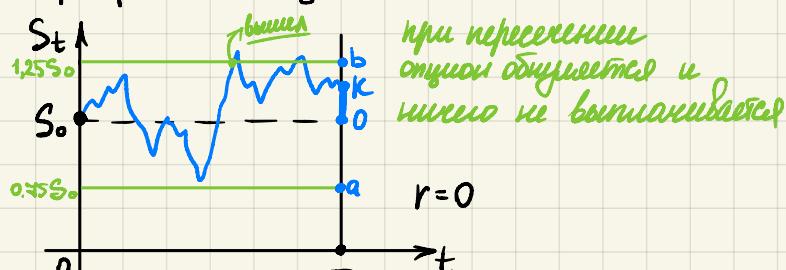
$$\Leftrightarrow \mathbb{E}[dh(X_t) | \Sigma_0] = \int_0^b p(x, t) dh(x_t)$$

$$\begin{aligned} \mathbb{E}[h(X_T)] &= \int_0^T \mathbb{E}[dh(X_t)] + h(X_0) \\ &\stackrel{\text{def}}{=} \int_0^T \left\{ \int_0^b p(u, T) h(u) du \right\} dt = \int_0^T \left[\int_0^b p(x, t) [h'(x), \mu(x, t) + \frac{1}{2} \sigma^2(x, t)] dx \right] dt = \\ &= h(X_0) + \int_0^T (I_1 + I_2) dt \quad \text{---} \end{aligned}$$

$$\begin{aligned} I_1 &= \int_0^b p(x, t) h'(x), \mu(x, t) dx = h(x) p(x, t), \mu(x, t) \Big|_0^b - \int_0^b h(x) \frac{\partial}{\partial x} (p(x, t), \mu(x, t)) dx \\ I_2 &= \frac{1}{2} \int_0^b p(x, t) h''(x) \sigma^2(x, t) dx = \frac{1}{2} [h'(x) p(x, t) \sigma^2(x, t) \Big|_0^b - \int_0^b h'(x) \frac{\partial}{\partial x} (p(x, t) \sigma^2(x, t)) dx] = \\ &= -\frac{1}{2} (h(x) \frac{\partial}{\partial x} (\dots) \Big|_0^b - \int_0^b h(x) \frac{\partial^2}{\partial x^2} (p(x, t) \sigma^2(x, t)) dx) \\ &= h(x_0) + \int_0^T \int_0^b \left(-h(x) \frac{\partial}{\partial x} (p, \mu) + \frac{1}{2} h(x) \frac{\partial^2}{\partial x^2} (p \sigma^2) \right) dx dt \Big| \frac{\partial}{\partial t} \\ &\int_0^b \frac{\partial p(u, T)}{\partial T} h(u) du = \int_0^b \left(-h(x) \frac{\partial}{\partial x} (p, \mu) + \frac{1}{2} h(x) \frac{\partial^2}{\partial x^2} (p \sigma^2) \right) dx = \\ &= \int_0^b h(x) \left\{ \frac{\partial p(x, T)}{\partial T} + \frac{\partial}{\partial x} (p(x, T) \mu(x, T)) - \frac{1}{2} \frac{\partial^2}{\partial x^2} (p(x, T) \sigma^2(x, T)) \right\} dx = 0 \end{aligned}$$

$$\begin{aligned} T \mapsto t \\ \begin{cases} \frac{\partial p}{\partial t} = -\frac{\partial}{\partial x} (\mu(x, t) p) + \frac{1}{2} \frac{\partial^2}{\partial x^2} (\sigma^2(x, t) p) \\ p(x, 0) = \delta(x - x_0) \end{cases} \quad (= \frac{\text{уравнение Фоккера-Планка}}{\text{уравнение Кошногорова}}) \\ L_1 p \quad L_1 = -\frac{\partial}{\partial x} (\mu \cdot) + \frac{1}{2} \frac{\partial^2}{\partial x^2} (\sigma^2 \cdot) \\ FP: \quad \frac{\partial p}{\partial t} = L_1 p \\ FK: \quad \frac{\partial f}{\partial t} = -L_2 f \quad L_1 = L_2^+ ! \end{aligned}$$

Барьерный опцион



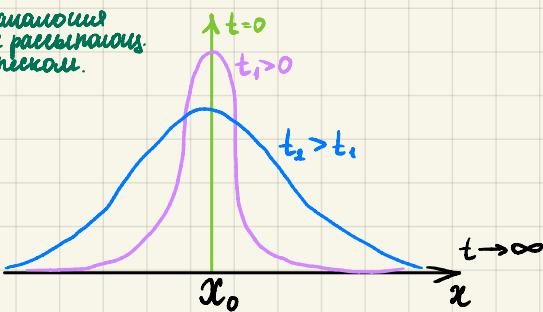
$$\begin{aligned} \text{up prob.} &= p(S_T \in (S_0, S_0 + dS) \mid \max_{0 \leq t \leq T} S_t \leq b \wedge \\ &\quad \min_{0 \leq t \leq T} S_t \geq a) \end{aligned}$$

$$\begin{aligned} P_{00} &= 0.01 K \quad \text{относит.} \\ F(x) &= \mathbb{P}\{ \text{Return} \leq x \} = \mathbb{P}\{ \frac{\text{Payoff}(T) - P_{00}}{P_{00}} \leq x \} = \mathbb{P}\{ \frac{\text{Payoff}(T)}{P_{00}} - 1 \leq x \} \\ &= \mathbb{P}\{ \frac{\text{Payoff}(T)}{P_{00}} \leq (1+x) P_{00} \} = \mathbb{P}\{ (S_T - K)^+ \leq y \} = \\ &\Rightarrow \begin{cases} (1) \quad y = 0 \quad x = -1 \\ (2) \quad y > 0 \end{cases} \end{aligned}$$

(1.8) - (1.10)

$$\frac{\partial p}{\partial t} = -\frac{\partial}{\partial x} (\mu p) + \frac{1}{2} \frac{\partial^2 p}{\partial x^2} (\sigma^2 p), \quad p(x, 0) = \delta(x - x_0)$$

аналогичные
рассыпывающиеся
некои.



$$\begin{aligned} M(x, t) &= 0 \\ \delta(x, t) &= 1 \\ dX_t &= dW_t \\ X_t &= X_0 + W_t \end{aligned}$$

$$p(x, t) = \frac{1}{\sqrt{2\pi t}} e^{-\frac{(x-x_0)^2}{2t}} \xrightarrow{t \rightarrow 0} \delta(x - x_0)$$

$$\frac{\partial p}{\partial t} = \frac{1}{2} \frac{\partial^2 p}{\partial x^2} - \text{уравнение броуновской}$$

$$\forall t > 0 \int_{-\infty}^{+\infty} p(x, t) dx = 1$$

$$\begin{aligned} Y_t &= f(X_t), \quad \frac{\partial f}{\partial t} = \frac{1}{6} \\ dY_t &= \tilde{\mu}(Y_t, t) dt + 1 \cdot dW_t \xrightarrow{\text{Girsanov Theorem}} dY_t = dW_t \end{aligned}$$

$$\begin{aligned} p(a, t) &= p(b, t) = 0 \quad a \leq x \leq b \\ p(x, 0) &= f(x - x_0) \end{aligned}$$

$$\Rightarrow p(S, T, a, b) dS = P[S_t \in (S, S+ds) \text{ } \& \max_{t \in [0, T]} S_t \leq b \text{ } \& \min_{t \in [0, T]} S_t \geq a]$$

$$\begin{aligned} X_t &= W_t \\ X_0 &= 0 \end{aligned}$$

$$\begin{aligned} p(x, 0) &= \delta(x) \\ p(\pm a, t) &= 0 \end{aligned}$$



$$\text{Ansatz } p(x, t) = p_1(x) p_2(t)$$

аналогичное
решение

$$p_1(x) \dot{p}_2(t) = \frac{1}{2} p_1''(x) p_2(t)$$

$$\frac{\dot{p}_2(t)}{p_2(t)} = \frac{1}{2} \frac{p_1''(x)}{p_1(x)} = \text{const} = \lambda$$

$$\begin{cases} p_1''(x) = 2\lambda p_1(x) \\ p_1''(x) - 2\lambda p_1(x) = 0 \\ p_1(\pm a) = 0 \end{cases}$$

$\lambda \geq 0$ и будем.
 $\Rightarrow \lambda < 0$

$$\begin{cases} p_1''(x) + \omega^2 p_1(x) = 0 \\ p_1(\pm a) = 0 \end{cases}$$

$$\begin{aligned} p_1(x) &= A \cos \omega x \\ \cos \omega a &= 0 \end{aligned}$$

$$\omega a = \frac{\pi}{2} + \pi n, \quad n = 0, 1, \dots$$

$$\omega = \omega_n = \frac{\pi}{a} \left(n + \frac{1}{2}\right)$$

$$\lambda = -\frac{\omega^2}{2} \Rightarrow \lambda_n = -\frac{\pi^2}{2a^2} \left(n + \frac{1}{2}\right)^2$$

$$\begin{aligned} p_1(x) &= A_n \cos \omega_n x \\ &= A_n \cos \left(\frac{\pi x}{a} \left(n + \frac{1}{2}\right) \right) \end{aligned}$$

$$\frac{\dot{p}_2(t)}{p_2(t)} = \lambda_n \Rightarrow p_2(t) = B_n e^{\lambda_n t} = B_n e^{-\frac{\pi^2}{2a^2} \left(n + \frac{1}{2}\right)^2 t}$$

$$C_n = B_n \cdot A_n \Rightarrow p_n(x, t) = C_n e^{-\frac{\pi^2 t}{2a^2} \left(n + \frac{1}{2}\right)^2} \cos \left(\frac{\pi x}{a} \left(n + \frac{1}{2}\right) \right)$$

$$p(x, t) = \sum_{n=0}^{\infty} p_n(x, t)$$

$$\sum_{n=0}^{\infty} C_n \cos \left(\frac{\pi x}{a} \left(n + \frac{1}{2}\right) \right) = \delta(x)$$

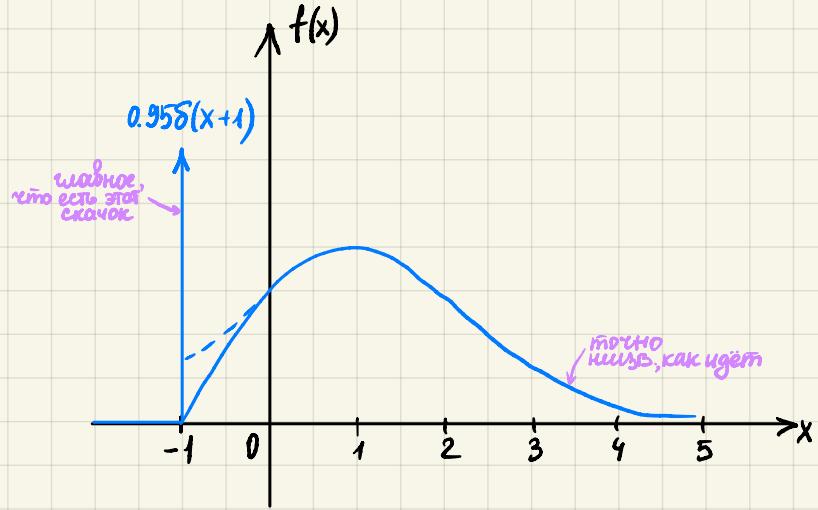
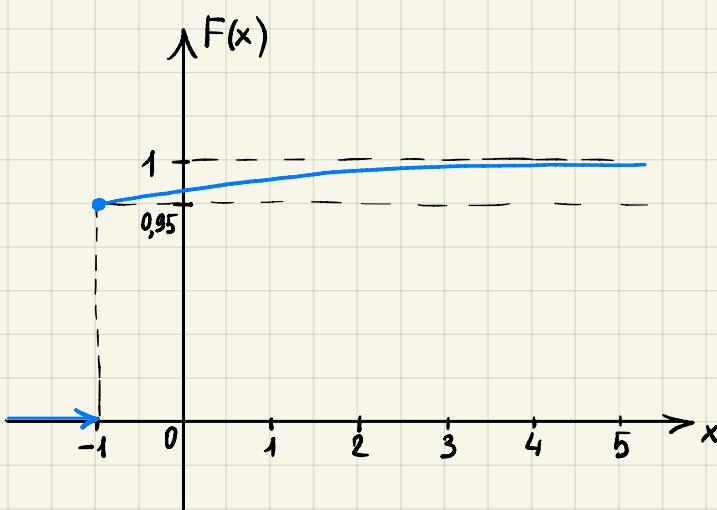
$$C_m \int_{-a}^{+a} \cos^2 \left(\frac{\pi x}{a} \left(m + \frac{1}{2}\right) \right) dx = \int_{-\infty}^{+\infty} \cos \left(\frac{\pi x}{a} \left(m + \frac{1}{2}\right) \right) \delta(x) dx \Rightarrow C_m = \frac{1}{a} \quad \forall m \geq 0$$

$$\begin{aligned} p(x, t, -a, a) &= \frac{1}{a} \sum_{n=0}^{+\infty} e^{-\frac{\pi^2}{2a^2}(n+\frac{1}{2})^2} \cos\left(\frac{\pi x}{a}(n+\frac{1}{2})\right) \frac{4a(-1)^n}{(2n+1)\pi} \\ \int_{-a}^a p(x, t, -a, a) dx &= \frac{1}{a} \sum_{n=0}^{+\infty} e^{-\frac{\pi^2}{2a^2}(n+\frac{1}{2})^2} \int_{-a}^a \cos\left(\frac{\pi x}{a}(n+\frac{1}{2})\right) dx = \frac{4}{\pi} \sum_{n=0}^{+\infty} \frac{(-1)^n}{2n+1} e^{-\frac{\pi^2}{2a^2}(n+\frac{1}{2})^2} \end{aligned}$$

$\begin{array}{l} t=0 \Rightarrow =1 \\ t>0 \Rightarrow <1 \end{array}$

$$p(s, T, a, b) = p_{tab}(s)$$

$$\begin{aligned} \mathbb{P}\{0 < \text{Payoff}(T) \leq (1+x)B_0\} &= \mathbb{P}\{K < S_T \leq (1+x)B_0 + K\} = \int_K^{(1+x)B_0} p_{tab}(s) ds \\ \mathbb{P}\{\text{Payoff}(T) \geq (1+x)B_0\} &= \int_{k+(1+x)B_0}^{+\infty} p_{tab}(s) ds \end{aligned}$$



$$\begin{array}{c} N \rightarrow \\ D(0,1) \\ D_{01}N \xrightarrow{-1} N \\ (1-D_{01})N \xrightarrow{-1} \text{Risk-free account} \\ \text{Barrier Option} \end{array}$$

$$D_{01} = e^{-rT}, \quad T = 1 \text{ year}, \quad r = 0.0425$$

$$\text{FP: } \frac{\partial p}{\partial t} = L_1 p, \quad L_1 = -\frac{\partial}{\partial x} (\mu \cdot) + \frac{1}{2} \frac{\partial^2}{\partial x^2} (\sigma^2 \cdot)$$

$$\text{FC: } \frac{\partial f}{\partial t} = -L_2 f, \quad L_2 = \mu \frac{\partial}{\partial x} + \frac{\sigma^2}{2} \frac{\partial^2}{\partial x^2}$$

$L_1 = L_2^+ - ?$ и compare aux. math.
и compare., a que bus math. это просто грамматон.

$$\begin{aligned} \langle L_1 p, f \rangle &= \langle p, L_2 f \rangle \\ \langle \tilde{x}, \tilde{y} \rangle &= \int_{-\infty}^{+\infty} \tilde{x}(x) \tilde{y}(x) dx \end{aligned}$$

$$\int_{-\infty}^{+\infty} L_1 p \cdot f dx \stackrel{?}{=} \int_{-\infty}^{+\infty} p \cdot L_2 f dx \quad \Leftrightarrow \int_{-\infty}^{+\infty} \frac{\partial p}{\partial t} f dx = - \int_{-\infty}^{+\infty} p \frac{\partial f}{\partial t} dx \Leftrightarrow$$

$$\int_{-\infty}^{+\infty} \frac{\partial}{\partial t} (p f) dx = 0 \Leftrightarrow \frac{\partial}{\partial t} \int_{-\infty}^{+\infty} p f dx = 0$$

$$\begin{aligned} f(x_t, t) &= \mathbb{E}[g(x_t) | \Sigma_t] \quad 0 \leq t \leq T \\ p(x, t) &: \text{PDF}(x_t) \end{aligned}$$

$$\begin{aligned} \int_{-\infty}^{+\infty} p f dx &= \text{const}(t) \quad \Leftrightarrow \mathbb{E}[f(x_t)] = \text{const}(t) \quad \Leftrightarrow \mathbb{E}[\mathbb{E}[g(x_t) | \Sigma_t]] = \text{const}(t) \\ &\Leftrightarrow \mathbb{E}[g(x_t)] = \text{const}(t)! \end{aligned}$$

$t=0$	t	$t+dt$	(3) Self-Financing	$t=T$
Money account $C_0 \cdot S_0$	M_t	$M_{t+dt} = M_t e^{r_f dt} \approx M_t (1 + r_f dt) = M_t + M_t r_f dt$	$M'_{t+dt} = M_{t+dt} - (\Delta_{t+dt} - \Delta_t) S_{t+dt}$	
Underlying $\Delta_0 \cdot S_0$	$\Delta_t \cdot S_t$	$\Delta_t S_{t+dt} = \Delta_t (S_t + dS_t) = \Delta_t (S_t + M_t r_f dt + \sigma_t dW_t)$	Rebalancing	
Option $-C_0$	$-C_t$	$-C_{t+dt} = f(S_t, t)$ <small>(1) Ansatz to be determined!</small>		$-C_T = -\varphi(S_T)$ Payoff (Europe)

где
наименее важные
части опциона не важны
свои изменения не важны

какие изменения
наиболее важные, но они
изменяются, то есть не важны

$$\Pi_0 = 0 \quad ; \quad \Pi_t = M_t + \Delta_t S_t - C_t$$

$$C_{t+dt} = C_t + dC_t = C_t + \frac{\partial f}{\partial t} + \frac{\partial f}{\partial S} (M_t r_f dt + \sigma_t dW_t) + \frac{1}{2} \frac{\partial^2 f}{\partial S^2} \sigma_t^2 dt = \\ = C_t + \left(\frac{\partial f}{\partial t} + \sqrt{M_t r_f dt + \sigma_t^2 \frac{\partial^2 f}{\partial S^2}} dt + \frac{\partial f}{\partial S} \sigma_t dW_t \right)$$

$$\begin{aligned} \Pi_{t+dt} &= M_t + M_t r_f dt + \Delta_t (S_t + M_t r_f dt + \sigma_t dW_t) - C_t - (\dots) dt - \frac{\partial f}{\partial S} \sigma_t dW_t = \\ &= \Pi_t + M_t r_f dt + \Delta_t M_t r_f dt + \Delta_t \sigma_t dW_t - (\dots) dt - \frac{\partial f}{\partial S} \sigma_t dW_t \end{aligned}$$

Δ_t max-at
ног наименее
коэффициент

$$d\Pi_t = [M_t r_f + \Delta_t M - \frac{\partial f}{\partial t} - M \frac{\partial^2 f}{\partial S^2} - \frac{\sigma_t^2 \partial^2 f}{2}] dt + \sigma_t (\Delta_t - \frac{\partial f}{\partial S}) dW_t$$

$\frac{\partial f}{\partial S} \Rightarrow \text{No Trend!}$

(2) Delta-Hedging

$$d\Pi_t = [M_t r_f - \frac{\partial f}{\partial t} - \frac{\sigma_t^2 \partial^2 f}{2}] dt = \Pi_t (e^{r_f dt} - 1) \approx \Pi_t r_f dt$$

(4) No arbitrage

$$M_t r_f - \frac{\partial f}{\partial t} - \frac{\sigma_t^2 \partial^2 f}{2} - \Pi_t r_f = (M_t - \frac{\partial f}{\partial S} S_t - C_t) r_f$$

$$-\frac{\partial f}{\partial t} - \frac{\sigma_t^2 \partial^2 f}{2} = r_f \frac{\partial f}{\partial S} S_t - r_f C_t$$

$$\frac{\partial f}{\partial t} + r_f \frac{\partial f}{\partial S} S_t + \frac{\sigma_t^2}{2} (S_t, t) \frac{\partial^2 f}{\partial S^2} - r_f C_t = 0$$

← +

Convective term Diffusion term Reactive term

(уп-нине Банка-Майнца-Мерсека)

→ SSM

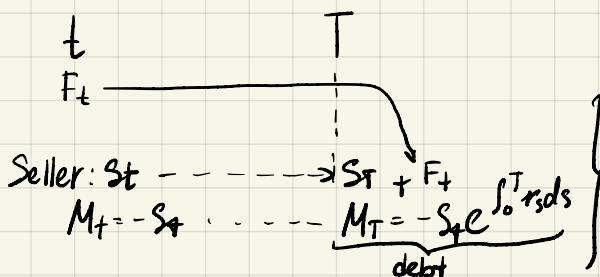
$$f(S_T, T) = \varphi(S_T) : \text{European Payoff}$$

Trivial Solutions:

- $f = S$
- $f = M_0 e^{\int_0^T r_s ds}$
- $f = B_T e^{-\int_0^T r_s ds}$

Stock
Money account
Bond

$F_t < S_T$: seller of loss
 $F_t > S_T$: buyer of loss



$$S_t - K e^{-\int_t^T r_s ds}$$

From call-like instrument?

$$f = F_t = S_t e^{\int_0^T r_s ds}$$

$$\frac{\partial f}{\partial t} = -r_f f$$

$$-r_f f + r_f \underbrace{S_t e^{\int_0^T r_s ds}}_f - r_f f = 0$$

Такое же
уравнение

C гибридизацией:

$$dS = (r_t - D_t)dt + \sigma_t dW_t$$

$$\begin{aligned} M_{t+dt} &= M_t e^{r_t dt} + \Delta_t S_t D_t dt \cong M_t (1 + r_t dt) + \Delta_t S_t D_t dt = M_t + (r_t M_t + \Delta_t S_t D_t) dt \\ \Delta_t S_t dt &= \Delta_t (S_t + dS_t) = \Delta_t (S_t + (M_t S_t D_t) dt + \sigma_t dW_t) = \Delta_t S_t + \Delta_t (M_t - D_t) dt + \Delta_t \sigma_t dW_t \\ C_{t+dt} &= C_t + dC_t = C_t + \frac{\partial f}{\partial t} dt + \frac{\partial f}{\partial S} (M_t - D_t) dt + \frac{\sigma^2}{2} \frac{\partial^2 f}{\partial S^2} dt \end{aligned}$$

$$\begin{aligned} \Pi_{t+dt} &= \Pi_t + (r_t M_t + \Delta_t S_t D_t) dt + \Delta_t (M_t - D_t) dt + \Delta_t \sigma_t dW_t - \frac{\partial f}{\partial t} dt - \frac{\partial f}{\partial S} ((M_t - D_t) M_t \\ &\quad + \sigma_t dW_t) - \frac{\sigma^2}{2} \frac{\partial^2 f}{\partial S^2} dt = \\ &= \Pi_t + [r_t M_t + \Delta_t S_t D_t + \Delta_t M_t - \Delta_t D_t - \frac{\partial f}{\partial S} M_t + \frac{\partial f}{\partial S} D_t - \frac{\sigma^2}{2} \frac{\partial^2 f}{\partial S^2}] dt + \\ &\quad + \sigma_t (\Delta_t + \frac{\partial f}{\partial S}) dW_t = \\ &= \Pi_t + [r_t M_t + \frac{\partial f}{\partial S} D_t S_t - \frac{\partial f}{\partial t} - \frac{\sigma^2}{2} \frac{\partial^2 f}{\partial S^2}] dt \stackrel{\text{No Arb}}{=} \Pi_t e^{r_t dt} \cong \Pi_t (1 - r_t dt) \end{aligned}$$

$$\begin{aligned} r_t M_t + \frac{\partial f}{\partial S} S_t D_t - \frac{\partial f}{\partial t} - \frac{\sigma^2}{2} \frac{\partial^2 f}{\partial S^2} &= \Pi_t r_t \\ \frac{\partial f}{\partial S} S_t D_t - \frac{\partial f}{\partial t} - \frac{\sigma^2}{2} \frac{\partial^2 f}{\partial S^2} &= r_t \frac{\partial f}{\partial S} S_t - r_t f \end{aligned}$$

$$\frac{\partial f}{\partial t} + [r_t - D_t] S \frac{\partial f}{\partial S} + \frac{\sigma^2 (S_t)}{2} \frac{\partial^2 f}{\partial S^2} - r_t f = 0 \quad \text{BSM with dividends}$$

from portfolio

$$f(S, T) = \varphi(S)$$

Stock — акции не уделяются.
Dividend Paying Stock: $f = S e^{r_t dt + D_t ds}$

Bond ✓
Money account ✓

Analytical solution to BSM PDE

Assume that $\delta(S, t) = \sigma_t S_t$

точно от времени

$$\frac{\partial f}{\partial t} + (r_t - D_t) S \frac{\partial f}{\partial S} + \frac{\sigma_t^2}{2} S^2 \frac{\partial^2 f}{\partial S^2} - r_t f = 0$$

$$(1) \quad \tilde{\tau} = T - t$$

(2) Eliminate the reactive term \Rightarrow Get Feynmann-Kac

$$\begin{aligned} (1) \quad f(S, t) &= f(S, T - \tilde{\tau}) = f_1(S, \tilde{\tau}) \\ (2) \quad f_1(S, \tilde{\tau}) &= f_1(S, \tilde{\tau}) e^{\int_0^{\tilde{\tau}} r_u du} \\ &\Rightarrow f_1(S, \tilde{\tau}) = f_2(S, \tilde{\tau}) e^{-\int_0^{\tilde{\tau}} r_u du} \end{aligned}$$

$$\frac{\partial f_1}{\partial \tilde{\tau}} = (r_{\tilde{\tau}} - D_{\tilde{\tau}}) S \frac{\partial f_1}{\partial S} + \frac{\sigma^2}{2} S^2 \frac{\partial^2 f_1}{\partial S^2} - f_1 r_{\tilde{\tau}}$$

$r_{\tilde{\tau}} = r_{T-\tilde{\tau}}, \quad \sigma_{\tilde{\tau}} = \sigma_{T-\tilde{\tau}}, \quad D_{\tilde{\tau}} = D_{T-\tilde{\tau}}$

$$f_2 e^{-\int_0^{\tilde{\tau}} r_u du} (-r_{\tilde{\tau}}) = (r_{\tilde{\tau}} - D_{\tilde{\tau}}) S \frac{\partial f_2}{\partial S} e^{-\int_0^{\tilde{\tau}} r_u du} + \frac{\partial f_2}{\partial \tilde{\tau}} e^{-\int_0^{\tilde{\tau}} r_u du} + \frac{\sigma^2}{2} S^2 \frac{\partial^2 f_2}{\partial S^2} e^{-\int_0^{\tilde{\tau}} r_u du} - D_{\tilde{\tau}} f_2 e^{-\int_0^{\tilde{\tau}} r_u du}$$

$$\frac{\partial f_2}{\partial \tilde{\tau}} = (r_{\tilde{\tau}} - D_{\tilde{\tau}}) S \frac{\partial f_2}{\partial S} + \frac{\sigma^2}{2} S^2$$

(3) Eliminate S, S^2 :

Работает только для норм. браузеров.

$$(3) \quad X = \ln \frac{S}{1} \quad S = 1 e^X$$

$$\frac{\partial f_2}{\partial X} = \frac{\partial f_2}{\partial S} \cdot \frac{dS}{dx} = \frac{\partial f_2}{\partial S} S$$

$$f_2(S, \tilde{\tau}) = f_2(e^X, \tilde{\tau}) = f_3(X, \tilde{\tau}) = f_3(\ln S, \tilde{\tau})$$

Запускание волатильности
очень медлено, т.к. если в стоке сильно
падают цены, то можно пренебречь
то убытки сток. ЧУП первого типа.
Диодн. цен "спасаются" спадами.
Также запускающее ($\Delta_t - \frac{\partial f}{\partial S}$) при
dWt перестает быть верным, а,
одновременно, и существует решение.
Волатильность — риск.

$$\begin{aligned} \frac{\partial^2 f_2}{\partial S^2} &= \frac{\partial}{\partial S} \left(\frac{\partial f_2}{\partial S} \right) = \frac{\partial}{\partial S} \left(\frac{1}{S} \cdot \frac{\partial f_3}{\partial X} \right) = \frac{\partial}{\partial X} \left(\frac{1}{S} \frac{\partial f_3}{\partial X} \right) \frac{dX}{dS} = \left(-\frac{1}{S^2} \frac{dS}{dX} \frac{\partial f_3}{\partial X} + \frac{1}{S} \frac{\partial^2 f_3}{\partial X^2} \right) \frac{1}{S} \\ &= \frac{1}{S^2} \left(\frac{\partial^2 f_3}{\partial X^2} - \frac{\partial f_3}{\partial X} \right) \Rightarrow S^2 \frac{\partial^2 f_2}{\partial S^2} = \frac{\partial^2 f_3}{\partial X^2} - \frac{\partial f_3}{\partial X} \\ \frac{\partial f_3}{\partial T} &= (P_T - D_T) \frac{\partial f_3}{\partial X} + \frac{\tilde{Z}_T^2}{2} \left(\frac{\partial^2 f_3}{\partial X^2} - \frac{\partial f_3}{\partial X} \right) \end{aligned}$$

(4) Eliminate the conv. term

$$\frac{\partial f_3}{\partial T} = \left(R_T - D_T - \frac{\tilde{Z}_T^2}{2} \right) \frac{\partial f_3}{\partial X} + \frac{\tilde{Z}_T^2}{2} \frac{\partial^2 f_3}{\partial X^2}$$

$$f_3(x, T) = f_3(f - \int_0^T \tilde{B}_u du, T) = f_4(f, T) \quad (4)$$

$$\frac{\partial f_4}{\partial f} = \frac{\partial f_3}{\partial X}, \quad \frac{\partial f_4}{\partial T} = \frac{\partial f_3}{\partial T} + \frac{\partial f_3}{\partial X} \frac{\partial X}{\partial T} = -\tilde{B}_T \frac{\partial f_3}{\partial X} + \frac{\partial f_3}{\partial T}$$

$$\frac{\partial f_3}{\partial T} = \frac{\partial f_4}{\partial T} + \tilde{B}_T \frac{\partial f_3}{\partial X} = \tilde{B}_T \frac{\partial f_3}{\partial X} + \frac{\tilde{Z}_T^2}{2} \frac{\partial^2 f_3}{\partial X^2} \Rightarrow \frac{\partial f_4}{\partial T} = \frac{\tilde{Z}_T^2}{2} \frac{\partial^2 f_4}{\partial f^2}$$

(5) Eliminate \tilde{Z}_T^2

$$y(T) = \int_0^T \tilde{Z}_u^2 du \Rightarrow T(y)$$

$$(5) f_4(f, T) = f_4(f, T(y)) = f_5(\xi, y)$$

$$\frac{\partial f_5}{\partial y} = \frac{\partial f_4}{\partial T} \frac{dT}{dy} = \frac{1}{\tilde{Z}_T^2} \frac{\partial f_4}{\partial T} \Rightarrow \frac{\partial f_5}{\partial T} = \tilde{Z}_T^2 \frac{\partial f_4}{\partial y}; \quad \frac{\partial f_4}{\partial y} = \frac{1}{2} \frac{\partial^2 f_4}{\partial f^2} - \text{точное упр}$$

$$f_5(\xi, y) = \frac{1}{\sqrt{2\pi y}} e^{-\frac{(\xi-a)^2}{2y}} \quad \forall a \in \mathbb{R}$$

$$f_5(\xi, y) = \frac{1}{\sqrt{2\pi y}} \int_{-\infty}^{+\infty} g(a) e^{-\frac{(\xi-a)^2}{2y}} da$$

$$f_5(\xi, y(T)) = f_4(f, T) = \frac{1}{\sqrt{2\pi y(T)}} \int_{-\infty}^{+\infty} g(a) e^{-\frac{(T-a)^2}{2y}} da \quad \begin{array}{l} \text{(ответ на чей-то вопрос)} \\ \text{смешаное сопряжение} \\ \text{обоснование erf-ти решетки} \\ \text{(лучше посмотреть 8 уроках)} \end{array}$$

Смысл динамо-аддитивизирующей - решений stocks на процессе, близкий к биржевому

$$\begin{aligned} \frac{\partial f_5}{\partial y} &= \frac{1}{2} \frac{\partial^2 f_5}{\partial \xi^2} \\ \frac{df_5}{dy} &= \frac{1}{2} (-\zeta \frac{\partial f_5}{\partial \xi}) F_5 \\ F_5(z, y) &= C(z) e^{-2\pi i z^2 y} \\ \downarrow \\ f_5(\xi, y) &= \int_0^{+\infty} e^{2\pi i z \xi} F_5(z, y) dz \\ &= \int_0^{+\infty} e^{2\pi i z \xi} \left[C(z) e^{-2\pi i z^2 y} \right] dz \\ &= \int_0^{+\infty} C(z) e^{2\pi i z(\xi - z^2 y)} dz \end{aligned}$$

$$\begin{aligned} z &\rightarrow z + iw \\ f_5(\xi, y) &= \int_{-\infty}^{+\infty} C(z) e^{2\pi i z(\xi - z^2 y)} dz \\ &= e^{\frac{i\xi^2}{2y} - 2\pi^2 z^2 y} \int_{-\infty}^{+\infty} C(z) e^{-2\pi^2 z^2 y} dz \\ &= 2\pi \xi i (z + iw) - 2\pi^2 (z + iw)^2 y \\ &= 2\pi \xi z - 2\pi \xi w - 2\pi^2 y (z^2 - w^2) - 4\pi^2 y w i \\ &= 2\pi \xi \underbrace{(z - 2\pi y w)}_{w = \frac{\xi}{2\pi y}} - 2\pi \xi w - 2\pi^2 y (z^2 - w^2) \\ &= -\frac{\xi^2}{2y} - 2\pi^2 z^2 y + \frac{\xi^2}{2y} = -\frac{\xi^2}{2y} - 2\pi^2 z^2 y \end{aligned}$$

(1.10) - (1.11)

$$\begin{aligned} dS_t &= (r_t - D_t) S_t dt \\ \frac{dS_t}{S_t} &= (r_t - D_t) dt \end{aligned}$$

Risk-neutral equities
FX

$$f_5(\xi, y(T)) = f_4(f, T) = \frac{1}{\sqrt{2\pi y(T)}} \int_{-\infty}^{+\infty} g(a) e^{-\frac{(f-a)^2}{2y(T)}} da$$

$$\text{Thus } y \rightarrow 0 \quad \frac{1}{\sqrt{2\pi y}} e^{-\frac{(f-a)^2}{2y}} \rightarrow \delta(a-f) \Rightarrow f_4(f, 0) = g(f)$$

Then yes. - результирует на ∞

$$f_3(x, T) = \frac{1}{\sqrt{2\pi y(T)}} \int_{-\infty}^{+\infty} g(a) e^{-\frac{(x + \int_0^T b_u du - a)^2}{2y(T)}} da$$

$$f_2(S, t) = \frac{1}{\sqrt{2\pi y(T)}} \int_{-\infty}^{+\infty} g(a) e^{-\frac{(\ln S + \int_0^T b_u du - a)^2}{2y(T)}} da$$

$$f_1(S, T) = \frac{e^{-\int_0^T r_u du}}{\sqrt{2\pi y(T)}} \int_{-\infty}^{+\infty} g(a) e^{-\frac{(\ln S + \int_0^T b_u du - a)^2}{2y(T)}} da$$

$$f(S, t) = \frac{e^{-\int_t^T r_u du}}{\sqrt{2\pi V(t)}} \int_{-\infty}^{+\infty} g(a) e^{-\frac{(\ln S + \int_t^T (r_u - D_u - \frac{\sigma^2}{2}) du - a)^2}{2V(t)}} da$$

$$f(y(T)) \Rightarrow V(t) = \int_t^T \sigma_u^2 du$$

$$V(t) = V_t = \int_t^T \sigma_u^2 du, \\ r_t - D_t - \frac{\sigma_t^2}{2} = b_t$$

По всем замечаниям компонент начальное условие: $f_5(\xi, 0) = \varphi(e^\xi)$

$$f_5(\xi, 0) = \varphi(e^\xi) = g(\xi) \Rightarrow g(a) = \varphi(e^a)$$

$$f(S, t) = \frac{e^{-\int_t^T r_u du}}{\sqrt{2\pi \int_t^T \sigma_u^2 du}} \int_{-\infty}^{+\infty} \varphi(e^a) e^{-\frac{(\ln S + \int_t^T (r_u - D_u - \frac{\sigma^2}{2}) du - a)^2}{2 \int_t^T \sigma_u^2 du}} da$$

относительно S_T
 $\frac{S_T - S_t}{(T-t)}$

$$e^a = Z, da = \frac{dz}{Z}$$

$$f(S, t) = \frac{e^{-\int_t^T r_u du}}{\sqrt{2\pi \int_t^T \sigma_u^2 du}} \int_0^{+\infty} \varphi(Z) \cdot e^{-\frac{(\ln \frac{S}{Z} + \int_t^T (r_u - D_u - \frac{\sigma^2}{2}) du)^2}{2 \int_t^T \sigma_u^2 du}} \frac{dz}{Z}$$

$$C_t = f(S_t, t)$$

European Call: $\varphi(z) = (z - k)^+ \quad z \geq k$

$$f_C(S, t) = \frac{D(t, T)}{\sqrt{2\pi \int_t^T \sigma_u^2 du}} \int_k^{+\infty} \left(1 - \frac{k}{Z}\right) e^{-\frac{(\ln \frac{S}{Z} + \beta)^2}{2\sigma^2(T-t)}} dz$$

$$r_t = r = \text{const}$$

$$D_t = D = \text{const}$$

$$\sigma_t = \sigma = \text{const}$$

$$\tau = T - t \quad (\text{time to expiration})$$

$$\beta = (r - D - \frac{\sigma^2}{2}) \tau$$

приложенные функции
сравнение
занести

$$X = \ln \frac{S}{Z} + \beta \\ dx = -\frac{dz}{Z}$$

$$f_C(S, t) = \frac{D(t, T)}{\sqrt{2\pi \int_t^T \sigma_u^2 du}} \int_{-\infty}^{\ln \frac{S}{Z} + \beta} (Se^{\beta - X} - k) e^{-\frac{X^2}{2\sigma^2(T-t)}} dx =$$

$$= \frac{Se^{-Dt}N_1 - Ke^{-rT}N_2}{N_{1,2} = QP \left(\frac{\ln \frac{S}{K} + (r-D \pm \frac{\sigma^2}{2})T}{\sigma\sqrt{T}} \right)}$$

$$QP(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-\frac{u^2}{2}} du$$

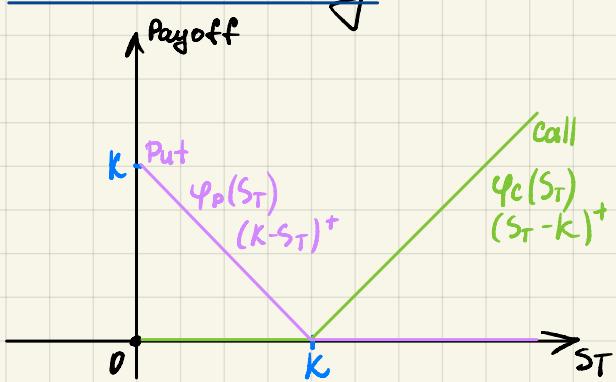
формула
Блэка-Шоудса-Мертонна

General Case:

$$\begin{aligned} (r - D \pm \frac{\sigma^2}{2})T &\longrightarrow \int_t^T (r_u - D_u \pm \frac{\sigma^2}{2}) du \\ e^{-DT} &\longrightarrow e^{-\int_t^T D_u du} \\ e^{-rT} &\longrightarrow e^{-\int_t^T r_u du} = D(t, T) \\ \sigma\sqrt{T} &\longrightarrow \sqrt{\int_t^T \sigma_u^2 du} \end{aligned}$$

$$t < T \quad S_t, -Ke^{-\int_t^T r_u du} \quad D(t, T)$$

Put-Call-Parity (Model-free)



$$\varphi_c(S_T) - \varphi_p(S_T) = S_T - K$$

- Long Call
- Short Put
- OR:
- Long Stock
- Short Bond
- OR:
- Borrow $K \cdot D(t, T)$

простейшие
рекомендации
(две схемы
безрискового
портфеля)

$$\text{Call_price}_t - \text{Put_price}_t = S_t - K D(t, T)$$

$$\frac{K}{T} \quad \frac{K}{T}$$

Диодифференциальные цены опционов ("Greeks")

$$\text{Call Px} = C(S_t, t, K, T, r, D)$$

некоторые
переменные
зависят
от контракта
и не
являются
бранными

самые
некоторые
переменные

$$\begin{aligned} t < T \\ \frac{\partial C}{\partial S} &= \Delta \\ 0 < \Delta < 1 \end{aligned}$$



$t=0$	t	Call: $\Delta = \frac{\partial C}{\partial S} = N_1 = \mathcal{P} \left(\frac{\ln \frac{S}{K} + (r - D - \frac{\sigma^2}{2})T}{\sigma \sqrt{T}} \right)$
$D = T-t = T$		$0 < \Delta < 1$
Money $C_0 = f_C(S_0, 0)$	M_t	$S > K : \Delta \rightarrow 1$
Hedge $\Delta_0 = \frac{\partial C}{\partial S} _0 \in (0, 1)$	$\Delta_t S_t$	$S < K : \Delta \rightarrow 0$
Option Pos. $-C_0$	$-C_t$	$S = K : \Delta \text{ not defined } (\rightarrow \frac{1}{2})$

$$\Pi_0 = 0$$

$$\Pi_t = 0$$

δ_{impl} : Mkt

$$\delta_{\text{mod}}^{\wedge} = \frac{1}{\sqrt{T}} \sqrt{\int_T^T \delta_{\text{mod}}^2(u) du}$$

$$S_0 = \$100 \rightsquigarrow S_T = \$120$$

(1) Buy stock $\rightsquigarrow \text{Ret} = 20\%$

(2) Buy option (call):

- at the money $K = 100$

$$C \approx 0.4 \frac{\delta_{\text{mod}}}{\sqrt{T}} K$$

$$C_0 = \$10 \Rightarrow \text{Payoff} = \$20$$

$$\text{PnL} = \$10$$

$$\text{Ret} = 100\%$$

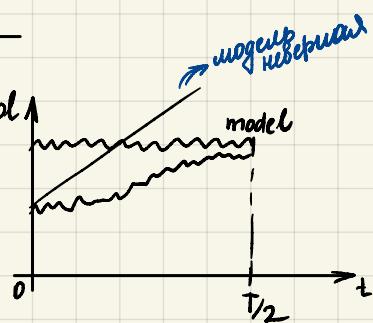
Barrier Opt:

$$B_0 = \$1 \Rightarrow \text{Payoff} = \$20$$

$$\text{PnL} = \$19$$

$$\text{Ret} = 1900\%$$

(3-5% cuyraged)



Использование
недвижимости
равновесия

Несоударственное
финансование