

Lecture 1

Risk Theory.

Var $\text{Var}_\lambda(X_T), \lambda \in (0, 1)$

Var - the λ quantile of the law of X_T with opposite sign

$$\text{Var}_\lambda(X_T) = -q_\lambda(X_T)$$

$$q_\lambda(X_T) = \inf\{x: P(X_T \leq x) \geq \lambda\}$$

$$\text{i.e. } F_{X_T}(q_\lambda) = \int_{-\infty}^{q_\lambda} p_{X_T}(z) dz \geq \lambda$$

If $X_T \geq 0$ then $q_\lambda \geq 0$ and $\text{Var}_\lambda < 0$
(there is no risk)

If $F_T(x) = P(X_T \leq x)$ the d.f. of X_T is cont. then $q_\lambda(x)$ is the solution to the equation $F_T(x) = \lambda$

Hermite - Ito - Wick polynomials

$$N(\bar{0}, B) \quad B = \begin{bmatrix} 1 & & p_{1j} \\ & \ddots & \\ p_{ij} & & 1 \end{bmatrix} \quad p_{ij} = \text{Cov}(X_i, X_j) \quad |p_{ij}| \leq 1$$

$$\text{if } \sum_{i,j \neq i} |p_{ij}| < 1$$

$$\text{then } \det B > 0 \quad B = \text{Cov}(X)$$

Lemma 1 If $\sum_{j \neq i} |p_{ij}| < \infty$

$$\text{and } D = \max_i \sum_{j \neq i} |p_{ij}| \quad (\text{BCL})$$

$$\|B - I\|_\infty < D$$

$$\vec{a} = \begin{bmatrix} a_1 \\ \vdots \\ a_n \end{bmatrix} \quad \|\vec{a}\|_\infty = \max_i |a_i|$$

$$\underbrace{\|B - I\|_\infty}_C \leq D \|a\|_\infty$$

$$\Rightarrow \|C\|_\infty = D$$

$$\|(Ca_i)\| = \left| \sum_j a_{ij} a_j \right| \leq \sum_j |a_{ij}| |a_j| \leq \|a\|_\infty \sum_{j \neq i} |a_{ij}| \leq D \|a\|_\infty$$

$$\|B - I\| < 1$$

$$B^{-1} = (I + (B - I))^{-1} = I + (B - I) + (B - I)^2 + \dots$$

$$\|B^{-1}\|_\infty \leq 1 + D + D^2 + \dots = \frac{1}{1-D}$$

$$\bar{X} = (x_1, \dots, x_d)$$

$$N(0, \bar{B})$$

after normalization

$$B = \begin{bmatrix} 1 & & \\ & \ddots & \\ & & 1 \end{bmatrix}$$

$d \approx 30-40$
размерности
тензора \tilde{X}

$$D = \max_i \sum_{j \neq i} |f_{ij}| \Rightarrow \det B > 0; \|B - I\|_\infty = D < 1$$

$$p_B(\bar{X}) = e^{-\frac{(B^{-1}\bar{X}, \bar{X})}{2}} \frac{(2\pi)^{d/2}}{\sqrt{\det B}}$$

$$\hat{p}_t(\bar{t}) = E e^{(\bar{t}, \bar{X})} = e^{-\frac{(B\bar{t}, \bar{t})}{2}}$$

$$X \sim N(0, 1) \quad d=1$$

$$E \varphi(x) = \int_{-\infty}^{+\infty} \varphi(x) p(x) dx$$

$$Y = \varphi(X), EY = 0, EY^2 < \infty$$

$$p(x) = \frac{1}{\sqrt{2\pi}} e^{-x^2/2}$$

$$(Y_1, Y_2) = EY, Y_2$$

$$L^2(\mu, p(x)) = \{ \varphi(x) : \int_{-\infty}^{\infty} \varphi p dx = 0, \int_{-\infty}^{\infty} \varphi^2 p dx = \|\varphi\|_2^2 \}$$

$$1, x, x^2, \dots, x^n$$

$$\int_{\mathbb{R}} x^{2k-1} p(x) dx = 0 \quad \int_{\mathbb{R}} x^{2k} p(x) dx = 1 \cdot 3 \cdot \dots \cdot (2k-1) = (2k-1)!!$$

Hermite's polynomials (Gaussians)

$$H_n(x), n = 0, 1, \dots$$

$$\int_{\mathbb{R}} H_n(x) H_m(x) p(x) dx = 0 \quad m \neq n$$

normalisation

$$\int_{\mathbb{R}} H_n^2 p(x) dx = 1 \quad \{H_n\} \text{ orthonom. basis}$$

$$H_n(x) = x^n + \dots$$

Нормализация и норм. ортонормированный

$$H_n(x), n = 0, 1, \dots$$

$$1, x, x^2, x^3 \in L^2(\mathbb{R}, p(x))$$

$$\begin{array}{l} \textcircled{1} 1 \perp x \\ \textcircled{2} 1 \perp \frac{x^2 - 1}{\sqrt{x^2 - 1}} \\ \textcircled{3} x^3 - c_1 x \perp \frac{x^2 - 1}{\sqrt{x^2 - 1}} \end{array}$$

$$\int_{\mathbb{R}} (x^2 - 1) p(x) dx = 0$$

$$\begin{aligned} \underline{Th 1} \quad e^{ax - a^2/2} &= 1 + (ax - a^2/2) + \frac{1}{2!} (ax - a^2/2)^2 + \dots \\ &= \sum_{n=0}^{\infty} \frac{a^n}{n!} H_n(x) \end{aligned}$$

$$H_n(x) = x^n + \dots$$

$$(H_n, H_n) = \int_{\mathbb{R}} H_n H_n p(x) dx = \delta_{n,n} \cdot n!$$

$$\int_{\mathbb{R}} (H_n/n!)^2 p(x) dx = 1$$

$$E \left(\sum_{h=0}^{\infty} \frac{a^h}{h!} H_h(x) \sum_{m=0}^{\infty} \frac{b^m}{m!} H_m(x) \right) = e^{ab}$$

$$E H_m(x) H_n(x) = 0, \quad m \neq n; \quad E H_n^2(x) = n!$$

ортонормированный

\Rightarrow ортонорм. система

The

$$a) H_{n+1}(x) = x \cdot H_n(x) - n H_{n-1}(x)$$

$$b) H'_n(x) = n H_{n-1}(x)$$

$$\triangleright e^{ax - a^2 x} = \sum_{n=0}^{\infty} \frac{a^n}{n!} H_n(x)$$

(a) def. H over q

(b) def. H over x

$L^2(\mathbb{R}, p(x))$

$\varphi(x), x \in \mathbb{R}$

$$\int_{\mathbb{R}} \varphi p dx = 0$$

$$\int_{\mathbb{R}} \varphi^2 p dx = \|\varphi\|_p^2 < \infty$$

$$H_n(x) = :X^n:$$

n^{th} Wick's degree

$$\left\{ \begin{array}{l} d/dx :X^n: = n :X^{n-1}: \\ d^n/dx^n :X^n: = n! \end{array} \right.$$

$$:X^0: = 1$$

$$:X^1: = x$$

$$:X^2: = x^2 - 1$$

$$E :X^n: :X^m: = \delta_{nm} n!$$

↑ raises. ann. + quad. form

Section 2

$$L^2(\Omega, P_0)$$

$$L^2(\Omega, P_I)$$

2 Hilb. space \mathcal{O} and \mathcal{I} are equivalent

$$\|y\|_B = \|y\|_I$$

срочно эквивалентны

$$C^- \|y\|_B \leq \|y\|_I \leq C^+ \|y\|_B$$

(отношение
вызвано
задан. между
конечн.)

$$B \neq I \Rightarrow L_B^2 \neq L_I^2$$

$$L^2(\Omega, P_B)$$

$$y = \varphi(\bar{x}) \quad \bar{x} \sim N(0, B)$$

$$E(y) = 0 \quad E(y^2) = \|y\|_B^2 < \infty$$

$$|y| \leq c$$

$$\cap L^2(\Omega, P_B)$$

$$\varphi(x_1, \dots, x_d) = \sum \left(\frac{x_1^{k_1}}{k_1!} \dots \frac{x_d^{k_d}}{k_d!} \right) \varphi_{k_1, \dots, k_d}$$

разложение по ортонорм. базису

всегда функцию можно разл. по лев. эрм.

$$E : x_1^{k_1} : \dots : x_d^{k_d} : = \begin{cases} 0, & k_1 + \dots + k_d \neq 1 \pmod{2} \\ \sum_{G \vdash \{1, \dots, d\}} (\prod f_{ij}) & , \text{ else} \end{cases}$$

$$x_i \sim N(0, 1)$$

$$e^{ax_1 - a^2/2} = \sum_{k=0}^{\infty} \frac{a^k}{k!} : x_1^k : \quad \uparrow \text{примит. групп.}$$

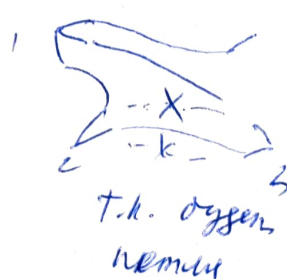
$$k_1=3$$

$$1 \leftarrow k_2=4 \Rightarrow a$$

$$2 \leftarrow k_2=1$$

канонич.
из нормировки

$$Eg_i$$



$$G_T = f_{14} f_{13} f_{12} f_{23}$$

$$k_1=3$$

$$k_2=2$$

$$k_3=2$$

$$k_4=1$$

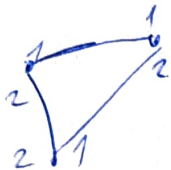
Eg. 1. $E: X_1^{k_1} \dots X_d^{k_d} = \begin{cases} 0 & k_1 \neq k_2 \\ \mu_j p_{k_1} & k_1 = k_2 \end{cases}$

$$B = \begin{bmatrix} 1 & \mu_j \\ \mu_j & 2 \end{bmatrix}$$



$$E: X_1^{k_1} X_2^{k_2} = \sum_{k_1, k_2} \mu_j p_{k_1}$$

Eg. 2. $X_1^2 X_2^2 X_3^2 = \beta_{12} \beta_{23} \beta_{13}$



$$e \beta_{12} a_1 a_2 + \beta_{23} a_1 a_3 + \beta_{23} a_2 a_3$$

$$= \dots \bigcirc \frac{a_1^2}{2!} \frac{a_2^2}{2!} \frac{a_3^2}{2!}$$

$$\dots \frac{(x+y+z)^3}{3!} = \frac{1}{3!} x! y! z! \frac{3!}{1!1!1!}$$

$$\dots e \frac{(\beta_{12} a_1 a_2 \dots)^3}{3!} \dots \frac{a_1^2}{2!} \frac{a_2^2}{2!} \frac{a_3^2}{2!} \cdot 1$$

Th. 9

$$\left| E: X_1^{k_1} \dots X_d^{k_d} \right| = \left| \sum_{\sigma \in (i, i')} \prod \mu_j p_{k_i} \right| \leq p^{N/2} \sqrt{k_1!} \dots \sqrt{k_d!}$$

$$\text{If } \rho = \max_i \sum_{j: i \neq j} |\beta_{ij}| < 1$$

$$E: \frac{X_1^{k_1}}{\sqrt{k_1!}} \dots \frac{X_d^{k_d}}{\sqrt{k_d!}} \leq \rho \frac{k_1 k_d}{2}$$

$$d=2 \quad \rho = \text{Cor}(k_1, k_2)$$

$$E: \frac{X_1^{k_1}}{\sqrt{k_1!}} \frac{X_2^{k_2}}{\sqrt{k_2!}} = \rho_{k_1, k_2} p^{k_1}$$

④ $S = S_1 + S_2$, где $S_1 = pS$, $S_2 = qS$, $p+q=1$

A) $S_1 (1 + 0.04 + 0.02 \xi_1)$

B) $S_2^{(1)} = \begin{cases} -1 & \text{w.p. } 1-10^{-4} \\ 1200 & \text{w.p. } 10^{-4} \end{cases}$

можно аппроксимировать

$$\tilde{S}_2 = S_B (1 + a_2 + b_2 \cdot \xi_2)$$

где $a_2 = 1200 \cdot \frac{1}{10^4} - 1 \cdot (1 - \frac{1}{10^4}) \approx -0.988$

$$b_2 = \sqrt{1200^2 \cdot \frac{1}{10^4} + 0.999 - 0.988^2} \approx 12$$

а) поскольку доходность второй стратегии в среднем ниже первой, то, очевидно, что для $\text{Max } \tilde{S}$ необходимо выбрать $p=1$, $q=0$

б) проверим вероятность убытка в случае выбора стратегии из п. а.)

$$P(S(1.04 + 0.02 \xi_1) < S) = P(\xi_1 < -2) = 0.02227 < 0.05$$

т.е. стратегия оптимальна при таких условиях.

Если бы доходность лотереи в среднем выше, чем доходность стратегии А,

тогда бы смысл использовать смешанную стратегию

Ⓘ

$$\begin{cases} F_{n+1} = F_n + F_{n-1} \\ F_0 = 1, F_1 = 1 \end{cases}$$

$$F_n = C_1 \lambda_1^n + C_2 \lambda_2^n$$

$$\begin{matrix} n=0 \\ n=1 \end{matrix} \quad \begin{cases} C_1 + C_2 = 1 \\ C_1 \lambda_1 + C_2 \lambda_2 = 1 \end{cases}$$

$$F_n = 2^n \quad \lambda^{n+1} = \lambda^n + \lambda^{n-1}$$

$$\lambda^2 - \lambda - 1 = 0$$

$$\lambda_{1,2} = \frac{1 \pm \sqrt{5}}{2}$$

↑ расч. решение

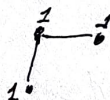
Ⓙ Теория перколяции

$$\mathbb{Z}^2 \quad x \in \mathbb{Z}^2$$

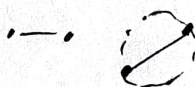
$$x = (x_1, x_2)$$

$$E(x) = \begin{cases} 1 & p \\ 0 & q=1-p \end{cases}$$

1 - связность



$\sqrt{2}$ - связность



port = 0,32
port = 0,66

океан 0 vs океан 1

Ⓜ

$$Y_{n+1} = \rho Y_n + \sigma \xi_n, \quad \xi_n \sim N(0,1)$$

$$\begin{aligned} Y_{n+1} &= \rho(\rho Y_{n-1} + \sigma \xi_{n-1}) + \sigma \xi_n = \dots = \\ &= \sigma \xi_n + \rho \sigma \xi_{n-1} + \rho^2 \sigma \xi_{n-2} + \dots \end{aligned}$$

$$\text{Var}(Y_{n+1}) = \sigma^2 (1 + \rho + \rho^2 + \dots) = \frac{\sigma}{1 - \rho^2}$$

$$E(Y_n, Y_{n-1}) = B(t)$$

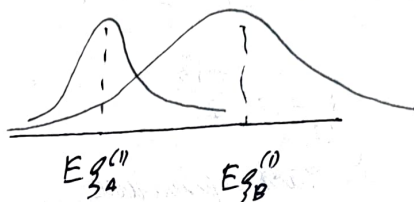
$$P(Y_n > h, Y_{n-1} > h, \dots, Y_{n-k+1} > h) \leq P\left\{\sum_{i=n-k+1}^n X_i > nh\right\}$$

IV D3 2 #1

$$\xi_A^{(1)} = \xi_A^{(0)} (1 + (0,04 + 0,01 \xi_A))$$

$$\xi_B^{(1)} = \xi_B^{(0)} (1 + (0,06 + 0,8 \xi_B))$$

$$\xi_1, \xi_2 \sim N(0,1)$$



Критерий? $A-p, B-(1-p)$

$$P(X > 10^4) = 0,95$$

$$\text{если } p=1 \Rightarrow P(\xi_A^{(1)} > 10^4) > 0,95$$

V D3 2 #2
 X_1, \dots, X_n н.о. $f(x), x \geq 0$

$$\{T=k\} = \delta(X_1 - X_2)$$

(Mosser)

$$EX_T = \max$$

$$P(X_T = M_n) = \max$$

$$M_n = \max(X_1, \dots, X_n)$$

$$X_1, \dots, X_n, EX_i < \infty$$

$$C_n = \max_T EX_T, C_n = F(C_{n-1})?$$

$$C_1 = EX_1 = a$$

$$C_2 = F(a) \quad C_3 = F(F(a))$$

$\rightarrow P(x), x \geq 0$



$\bullet x_i > h_n \text{ stop } T=1$

$$\left(\frac{p(x)}{\int_{h_n}^{\infty} p(x) dx} \cdot I(x) \right)_{[h_n, \infty)} \quad \bullet \int_{h_n}^{\infty} p(x) dx \leq 1$$

$P(h_n)$

$$E(X | X > h_n) = \int_{h_n}^{\infty} x p(x) dx / \int_{h_n}^{\infty} p(x) dx$$

(IX) $[R_{32} \neq 4]$

$$\bar{X}_{n+1} = A \bar{X}_n$$

$$\bar{X}_n = \lambda^n \vec{\xi}$$

$$(A - \lambda I) \vec{\xi} = 0$$

$$\lambda_{\text{ap-e}} \quad \gamma_{\text{p-e}}$$

$$\text{Thm. } \lambda_{\max} \text{ берётся } (\lambda > 0)$$

(I) $[R_{32} \neq 1] \quad S = 10^4$

$$x_1, x_2, x_3; \quad x_1 + x_2 + x_3 = \sqrt{S}$$

$$p_1 + p_2 + p_3 = 1$$

$$\tilde{x}_1 = x_1 (1 + a_1 + b_1 \xi_1)$$

$$\tilde{x}_2 = (x_2 (1 + a_2 + b_2 \xi_2))$$

$$\tilde{x}_3 = \dots$$

$$a_1 = 0.04$$

$$a_2 = 0.06$$

$$a_3 = 0.1$$

$$b_1 = 0.01 \quad \xi_1 \sim N(0,1)$$

$$b_2 = 0.1$$

$$b_3 = 0.3$$

$$P \{ \tilde{x}_1 + \tilde{x}_2 + x_3 < S \}$$

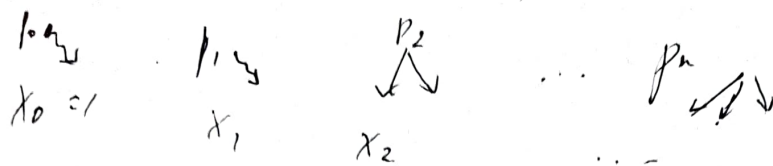
$$\tilde{x}_1 + \tilde{x}_2 + \tilde{x}_3 = x_1(\dots) + \dots \sim N$$

$$P \{ \underbrace{a_1 x_1 + a_2 x_2 + a_3 x_3}_{>0} + b_1 x_2 \xi_2 + \dots < 0 \}$$



IV

(XV) Bernoulli process (Gatton - Watson)



$$X_{n+1} = \xi_1 + \dots + \xi_n$$

ξ_i i.i.d.

$$\xi : \Psi_\xi(t) = E e^{it\xi} = \int_{\mathbb{R}} e^{itx} p(x) dx \quad |p_\xi(0)| = 1$$

$$\xi > 0 \quad \Psi_\xi(t) = E e^{-t\xi} = \int_0^\infty e^{-tx} p(x) dx$$

$$\xi \in [a, \dots, n] \quad \varphi(z) = E z^\xi = p_0 + p_1 z + \dots + p_n z^n + \dots, \quad |z| \leq 1$$

$$(1) \quad \begin{matrix} 0 & 1 & \dots & n & \dots \\ p_0 & p_1 & & p_n & \dots \end{matrix}$$

$$(2) \quad F_V(z) = E(z^V) = p_0 + p_1 z + \dots + p_n z^n$$

$$(3) \quad \begin{matrix} 0 & 1 & \dots \\ p_0 & p_1 & \dots \end{matrix}$$

$$E z^{\xi_i} = \Phi_\xi(z)$$

$$S = \xi_1 + \dots + \xi_n$$

$$E z^S = E z^{\xi_1 + \dots + \xi_n} = F_V(\Phi_\xi(z))$$



$$X_i < h_n \quad p(X_i < h_n) = \int_0^{h_n} p(x) dx$$

$$C_n = \max_{h_n} \left\{ p(X_i > h_n) \cdot E(X_i | X_i > h_n) + p(X_i < h_n) \cdot C_{n-1} \right\}$$

$$= \max_{h_n} \left\{ \int_{h_n}^{\infty} x p(x) dx + \int_0^{h_n} p(x) dx \cdot C_{n-1} \right\}$$

$$f'(x) = 0$$

$$E(X_i | X_i > h_n) = \frac{\int_{h_n}^{\infty} x p(x) dx}{\int_{h_n}^{\infty} p(x) dx}$$

$$-h_n \cdot p(h_n) + C_{n-1} \cdot p(h_n) = 0$$

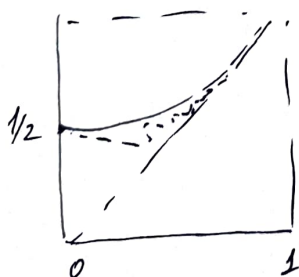
$$\boxed{C_{n-1} = h_n}$$

$$C_n = \int_{C_{n-1}}^{\infty} x p(x) dx + C_{n-1} \int_0^{C_{n-1}} p(x) dx \quad (\Rightarrow) C_n = F(C_{n-1})$$

$$p(x) = \begin{cases} 1 & x \in (0, 1) \\ 0 & x \notin (0, 1) \end{cases}$$

$$C_n = \frac{1 - C_{n-1}^2}{2} + C_{n-1}^2 = \frac{1 + C_{n-1}^2}{2}$$

$$C_n = f(C_{n-1}); \quad f(x) = \frac{1+x^2}{2}$$



$$C_1 = 1/2 \quad C_2 = \frac{1 + (1/2)^2}{2} = \frac{5}{8}$$

$$C_n = 1 - \frac{C_0}{n^\alpha} + \frac{C_1}{n^{\alpha+1}} + \dots$$

$$\left| \frac{1}{(n+1)^\alpha} = \frac{1}{n^\alpha} \left(1 + \frac{1}{n} \right)^{-\alpha} \right|$$

$$= 1 - \frac{C_0}{(n+1)^\alpha} \dots = \frac{1}{2} + \frac{1}{2} \left(1 + \frac{C_0}{n^\alpha} + \dots \right)^2 \quad (\Rightarrow) C_n = 1 - \frac{2}{n} + o(1/n^2)$$

$$E \text{Max}(X_1, \dots, X_n) = \tilde{c}_n$$

$$P(X_n \leq x) = F(x) = \int_0^x p(x) dx$$

$$M_n = \max(X_1, \dots, X_n)$$

$$F_{M_n}(x) = P(\max(X_1, \dots, X_n) \leq a) \stackrel{!!}{=} P(X_1 \leq a, \dots, X_n \leq a) =$$

$$= P(X_1 \leq a)^n = F(a)^n$$

$$E M_n = n \int_0^\infty a p(a) F^{n-1}(a) da \quad p_{M_n}(x) = n F^{n-1}(a) p(a)$$

e.g. $X_1 \sim \text{Unif}([0, 1])$

$$F_{M_n}(x) = F^n(a) = a^n$$

$$p_{M_n}(x) = n a^{n-1}$$

$$E M_n = \int_0^1 a p_{M_n}(a) da = n \int_0^1 a^n da = \frac{n}{n+1} = 1 - \frac{1}{n+1} =$$

$$E M_n = 1 - \frac{1}{n} + \dots$$

$$= 1 - \frac{1}{n} + O(1/n^2)$$

$$E X_n = 1 - \frac{2}{n} + \dots$$

$$\frac{1 - E X_n}{1 - E M_n} = 2$$