

A. A. (1)

$$\alpha = (1 \ 0 \dots 0) \quad m - \text{active}$$

$$\alpha = (\alpha_1 \ \alpha_2 \ \dots \ \alpha_n) \quad m+1 = \emptyset$$

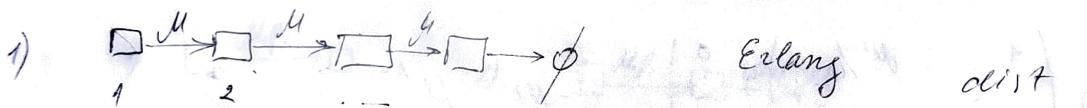
$$e^{Qt} = I + Qt + \frac{1}{2!} Q^2 t^2 + \dots$$

$$\det [e^{Qt}] = e^{t \cdot \text{tr}[Q] \cdot t}$$

$$Q = \begin{pmatrix} -\mu_1 & \mu_1 & & \\ & -\mu_2 & \mu_2 & \\ & & \ddots & \\ & & & -\mu_n & \mu_n \\ 0 & 0 & \dots & 0 \end{pmatrix} \quad \sum_j \mu_{ij} = 0 \quad \forall i$$

$$Q = \begin{pmatrix} T & t \\ 0 & 0 & 0 \end{pmatrix} \quad \boxed{t = -T \neq 0}$$

↑ mode замкнутости



$$1) \quad \left(\begin{array}{cc|c} -\mu & \mu & & 0 & | & 0 \\ 0 & -\mu & \mu & & | & : \\ \vdots & & & & | & : \\ 0 & & & -\mu & \mu & -\mu \\ 0 & & & & & \mu \end{array} \right) \quad t = \begin{pmatrix} 0 \\ \vdots \\ 0 \\ \mu \end{pmatrix}$$

$$T = -\mu I + \mu B, \quad \text{ze} \quad B = \begin{pmatrix} 0 & 1 & 0 \\ 0 & \ddots & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

1

$n \times n$

$$e^{Tx} = e^{-\mu Ix} \cdot e^{\mu B}$$

$$B^2 = \begin{pmatrix} 0 & 0 & 1 & 0 \\ 0 & \ddots & 1 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

$$e^{\mu B} = I + \mu B + \frac{\mu^2}{2!} B^2 + \dots \quad B^n = 0$$

$$e^{\mu B} = \begin{pmatrix} 1 & \mu & \mu^2/2! & \dots & \mu^{n-1}/(n-1)! \\ & \ddots & \ddots & \ddots & \vdots \\ & & \ddots & \ddots & \mu^{n-1}/(n-1)! \\ 0 & & & & \mu \\ & & & & 1 \end{pmatrix}$$

$$f(x) = x \cdot e^{Tx} \cdot t \quad - \text{nuomo orno}$$

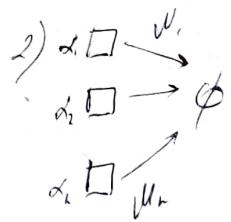
$$(1 \ 0 \ \dots \ 0) \begin{pmatrix} 1 & \mu & \mu^2/2! & \dots & \mu^{n-1}/(n-1)! \\ & \ddots & \ddots & \ddots & \vdots \\ & & \ddots & \ddots & \mu^{n-1}/(n-1)! \\ 0 & & & & 1 \end{pmatrix} \begin{pmatrix} 0 \\ \vdots \\ 0 \\ \mu \end{pmatrix} = e^{-\mu x} \cdot \frac{\mu^{n-1}}{(n-1)!} \quad \mu =$$

$$= \frac{\mu^n}{(n-1)!} e^{-\mu x}$$

Erlang dist

$$1 - F(x) = P(\xi > x) = \sum_{j=1}^n \alpha_j (e^{Tx})_{ij} = x e^{Tx} \mathbf{1}_{n,1}$$

$$S(\alpha) = \alpha e^{Tx} \underset{q}{\mathbf{1}_{n,1}} \quad f(x) = -S'(x) = -\alpha e^{Tx} \cdot \underbrace{T \cdot \mathbf{1}}_{-t} = \alpha e^{Tx} t$$



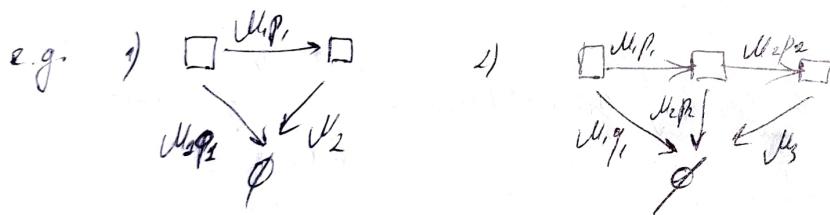
$$f(x) = \sum_{j=1}^n \alpha_j \mu_j e^{-\mu_j x}$$

$$\alpha = \left(\begin{array}{cccc|c} -\mu_1 & 0 & \dots & & \mu_1 \\ & -\mu_2 & 0 & \dots & \mu_2 \\ & & \ddots & & \vdots \\ & & & -\mu_n & \mu_n \\ 0 & \dots & & 0 & 0 \end{array} \right)$$

$$f(x) = (\alpha_1 \dots \alpha_n) \begin{pmatrix} e^{-\mu_1 x} & \dots & 0 & | & \mu_1 \\ \vdots & \ddots & \vdots & | & \vdots \\ 0 & \dots & e^{-\mu_n x} & | & \mu_n \end{pmatrix}$$

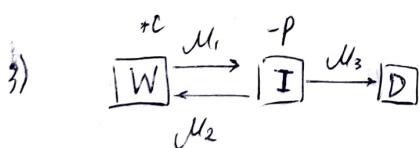
3) $\square \xrightarrow{\mu_1} \square \xrightarrow{\square} \square \xrightarrow{\dots} \square$
 $\mu_1 \downarrow \quad \downarrow \quad \downarrow \quad \downarrow \quad \downarrow$
 ϕ

$$\alpha = \left(\begin{array}{ccccc|c} -\mu_1 & \mu_1 p_1 & \dots & & & \alpha_1 q_1 \\ & -\mu_2 & \dots & & & \\ & & \ddots & & & \\ & & & \ddots & & \\ 0 & & & & \ddots & 0 \\ & & & & & 0 \end{array} \right) \quad p_1 + q_1 = 1$$



$$\mu_1 \neq \mu_2$$

$$\mu_1 \neq \mu_2 \neq \mu_3$$



$$\begin{matrix} W & \left(\begin{array}{cc|c} -\mu_1 & \mu_1 & 0 \\ \mu_2 & -\mu_1 - \mu_2 & \mu_3 \\ \hline 0 & \dots & 0 \end{array} \right) \\ I & \\ D & \end{matrix}$$

$$f(x) = (\alpha_1 \dots \alpha_2) \begin{pmatrix} e^{-\mu_1 x} & e^{\mu_1 x} & 0 \\ e^{\mu_2 x} & e^{(\mu_2 - \mu_1)x} & \mu_3 \end{pmatrix}$$

$$-C(-T)_{11} + P(-T)_{12} = 0$$

1. wyr. wyp. 1
2. I-T

$$= \begin{pmatrix} \alpha_1 e^{-\mu_1 x} + \alpha_2 \cdot e^{\mu_1 x} & 0 \\ \alpha_1 e^{\mu_2 x} + \alpha_2 \cdot e^{(\mu_2 - \mu_1)x} & \mu_3 \end{pmatrix} =$$

$$= \alpha_1 \mu_3 e^{\mu_1 x} + \alpha_2 e^{(\mu_1 - \mu_2)x} \cdot \frac{\alpha_2}{\alpha_1} \quad 2x1$$

Laplace Transform

$$\int_0^\infty e^{-sx} \alpha e^{tx} dt = \alpha(sI - T)^{-1}$$

$$\alpha \int_0^\infty e^{-sTx} e^{tx} dt = \alpha (sI - T)^{-1}$$

$$E[\xi] = -(-1) \alpha (sI - T)^{-2} t \Big|_{s=0} =$$

$$= \alpha (-T)^{-2} t = -\alpha (-T)^{-1} \cdot 1$$

$$-Tt = 1$$

$$E[\xi^k] = (-1)^k k$$

$$\alpha (sI - T)^{-k-1} t \Big|_{s=0} = (-1)^{k+1} k! \alpha (-T)^{-k} (-T)^{-1} (-T)^{-1}$$

$$= (-1)^k k! \alpha (-T)^{-k} 1$$

time
↓

$$E_{ij} = E_i [\overset{\sim}{\alpha_j}] = (-T)_{ij}$$

Corresponding rows and columns go now together

(beginning w/ i)

$$E_{ij} = \sum_{k \neq j} \frac{T_{ik}}{T_{ii} - T_{kk}} E_{kj} \quad (\Rightarrow E(-T)^{-1} = I) \quad \begin{pmatrix} T_{11} & T_{12} & \dots & T_{1n} \\ \vdots & \ddots & \ddots & \end{pmatrix}$$

$$E_{ii} = -\frac{1}{T_{ii}} + \sum_{k \neq i} \frac{T_{ik}}{-T_{ii}} E_{ki} \quad \Rightarrow \sum_k E_{ik} E_{kj} = 0 \quad i \neq j$$

$$\Rightarrow (-T_{ii}) E_{ij} + \sum_{k \neq i} (-T)_{ik} E_{kj} = 0$$

$$-\sum_k E_{ik} T_{ki} = 1$$

$$\Rightarrow (-T_{ii}) E_{ii} + \sum_{k \neq i} (-T)_{ik} E_{ki} = 1$$

$$Q = \begin{pmatrix} -3/12 & 9/14 & 8/7 \\ 9/2 & -1/2 & 2 \\ 0 & 0 & 0 \end{pmatrix} \quad \alpha = \begin{pmatrix} 1/2 & 1/2 \end{pmatrix}$$

$$\lambda_1 = -1 \quad \lambda_2 = -6$$

$$\lambda_1 + \lambda_2 = \text{tr}(Q)$$

$$\lambda_1 \cdot \lambda_2 = \det(Q)$$

np. uob. codom. bewegen:

$$\pi = Q(t_{21}, t_1 - t_n) = \alpha \left(\frac{7}{2}, \frac{1}{2} \right)$$

$$v = f \left(\frac{t_{12}}{t_1 - t_n} \right) = f \left(\frac{9/24}{9/2} \right)$$

$$\langle \pi, v \rangle = 1$$

$$\Leftrightarrow ab = 2/5$$

$$P_1 = \begin{pmatrix} \pi_1 v_1 & \pi_2 v_1 \\ \pi_1 v_2 & \pi_2 v_2 \end{pmatrix} \quad - \text{normalprojektion} \Rightarrow \begin{pmatrix} 9/10 & 9/20 \\ 7/10 & 1/10 \end{pmatrix}$$

$$P_2 = \begin{pmatrix} \pi_2 v_2 & -\pi_2 v_1 \\ -\pi_1 v_2 & \pi_1 v_1 \end{pmatrix} \quad I = P_1 + P_2 \quad P_1^2 = P_2^2 = I$$

$$P_1 P_2 = P_2 P_1 = 0$$

$$Q = \lambda_1 P_1 + \lambda_2 P_2$$

$$f(Q) = f(\lambda_1) \cdot P_1 + f(\lambda_2) \cdot P_2 \quad T^{-k} = (-1)^k (T_1 + e^{-k} T_2)$$

$$E[f^k] = (-1)^k k! \propto (-T)^{-k} \mathbb{1}$$

$$f(x) = \alpha e^{Tx} t \quad e^{Tx} = e^{-x} \begin{pmatrix} 9/10 & 9/10 \\ 7/10 & 1/10 \end{pmatrix} + e^{-6x} \begin{pmatrix} \cdot & \cdot \\ \cdot & \cdot \end{pmatrix}$$

разложение в степенях

$$e^{Ax} = I + \underbrace{\lambda_1 T_1 + \lambda_2 T_2}_A + \frac{1}{2} (\lambda_1 T_1 + \lambda_2 T_2)^2 + \dots = e^{\lambda_1 x} T_1 + e^{\lambda_2 x} T_2$$

$$\underbrace{\lambda_1^2 T_1^2}_I + \underbrace{2\lambda_1 \lambda_2 T_1 T_2}_0 + \underbrace{\lambda_2^2 T_2^2}_0$$

* неупорядоченное пред. неупорядоченное нн

$x \sim (\alpha, T_1)$ — gray norma (PH-representation)

$y \sim (\beta, T_2)$

$\min [x, y] \sim (\alpha \otimes \beta, T_1 \oplus T_2)$

$\max [x, y] \sim \text{clerk}(\alpha \otimes \beta, 0, 0) \begin{pmatrix} T_1 \oplus T_2 & I \oplus t^{(1)} & t^{(2)} \otimes I \\ 0 & T_1 & 0 \\ 0 & 0 & T_2 \end{pmatrix}$
 $(mn + n + m)_x / (mn + n + m)$

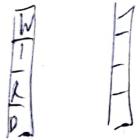
$\sum_{j=1}^n x_j \sim (\alpha \otimes \beta, T \otimes I + t \alpha \otimes S)$

$x \sim (\alpha, T)$ PH

$N \sim (\beta, S)$ (PH) group

Tiele equation

$A \quad W \quad \text{of more}$



$\circ \quad \vdots \quad \vdots$

$$\frac{P\ddot{a}_{x:t} - S A_{x:t}^1}{t P_x v t} = S A_{x+t:n-t} - P\ddot{a}_{x+t:n-t}$$

$t P_x v t$

actuarial accumulation

ann. insurance

④ claim $\mu_{21} \neq 0$ (8×8)

go 1 cu. 4 x 4

go 2 cu. 8 x 8
(2x2+1x2)

* 5.3.20

Maymunlar



$$p = 1/2$$

$$p = 1/2 + 3/1000$$

$$p(A \text{ min}) = 10\%$$

$$\approx 0,95$$

$$p(B \text{ min}) = 1/100$$

$$\approx 0,05$$

$$S_n = X_1 + \dots + X_n$$

$$P = 1/2 \Rightarrow S_n - \text{martingale} \quad E(S_{n+1} | F_n) = S_n \Rightarrow 10\%, 1\%$$

$$\left(\frac{q}{p}\right) S_n - \text{martingale?}$$

\hookrightarrow Odm. no meospenn gýda $\Rightarrow P_s \cup P_c$

$$E X_2 = E X_0$$

$$t B(t) - 1/3 B^3(t) - \text{martingale?}$$

gamma-process

neg. gauss process

brown. pr.

$$E e^{rX} = (1+r) E X$$

$$x(R) = 0$$

$$e^{sX(t)} - \text{musp. kenges}$$

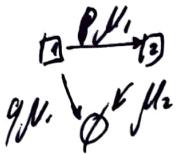
$$e^{sX(t) - R(s)t} \leftarrow \text{Bergen maymunlari}$$

$$e^{R(X(t))} - \text{martingale } (s=0)$$

$$e^{\sigma B(T(t))} \xrightarrow{\text{cydysymmetri}} \text{odan. mopsob (manomusimli)}$$

a n B

A.1 (2)



$$1) I = T_1 + T_2 \Rightarrow T_1 T_2 = T_2 T_1 = 0$$

$$2) T_1^2 = T_1 \quad T_2^2 = T_2$$

$$T = \begin{pmatrix} -\mu_1 & P\mu_1 \\ 0 & \mu_2 \end{pmatrix} \begin{pmatrix} P\mu_2 \\ \mu_2 \end{pmatrix}$$

$$T = -\mu_2 \begin{pmatrix} 0 & P\mu_1/(\mu_2 - \mu_1) \\ 0 & 1 \end{pmatrix} - \mu_1 \begin{pmatrix} 1 & -P\mu_2/(\mu_1 - \mu_2) \\ 0 & 0 \end{pmatrix}$$

2 pair of linear equations

$$\pi = a(t_0, \dots) = a(0, \mu_1 - \mu_2)$$

$$ab = \frac{1}{(\mu_2 - \mu_1)^2}$$

$$v = b(t_0) = b(P\mu_1 / (\mu_1 - \mu_2))$$

$$T_1 = \begin{pmatrix} \pi_1 v_1 & \pi_2 v_1 \\ \pi_1 v_2 & \pi_2 v_2 \end{pmatrix}$$

$$T_2 = \begin{pmatrix} \pi_2 v_2 & -\pi_2 v_1 \\ -\pi_1 v_2 & \pi_1 v_1 \end{pmatrix}$$

$$e^{Tx} = e^{-\frac{\mu_2}{2}x} \underbrace{T_1}_{''} + e^{\frac{\mu_1}{2}x} \underbrace{T_2}_{''} = \dots$$

$$f(x) = (1, 0) \begin{pmatrix} e^{-\mu_1 x} (e^{-\mu_2 x} - e^{\mu_2 x}) \frac{P\mu_1}{\mu_1 - \mu_2} \\ 0 \end{pmatrix} \begin{pmatrix} P\mu_1/(\mu_1 - \mu_2) \\ 1 \end{pmatrix} \begin{pmatrix} 1 & -P\mu_1/(\mu_1 - \mu_2) \\ 0 & 0 \end{pmatrix}$$

$$f(x) = e^{-\mu_1 x} P\mu_1 + \frac{P\mu_1 \mu_2}{\mu_2 - \mu_1} (e^{-\mu_1 x} - e^{\mu_1 x})$$

$$(1) f(x) = -P\mu_1 e^{-\mu_1 x} (1 + u_2) \begin{pmatrix} P\mu_1 \\ u_2 = \mu_2 \end{pmatrix}$$

Пусть $\mu_1 = \mu_2 = \mu$

$$\mu e^{-\mu x} \left(1 - \frac{e^{-\Delta \mu x}}{-\Delta \mu} - 1 \right) = \underbrace{\mu e^{-\mu x} (q + p_{\mu x})}_{0/0}$$

но логично

$$\exp \begin{pmatrix} -\mu & p_\mu \\ 0 & \mu \end{pmatrix} = \begin{pmatrix} e^{-\mu x} & e^{-\mu x} (a + b x) \\ 0 & e^{-\mu x} \end{pmatrix} \begin{pmatrix} \mu^2 \\ 0 \end{pmatrix} = \begin{pmatrix} e^{-\mu x} & -a e^{-\mu x} - b x e^{-\mu x} \\ 0 & e^{-\mu x} \end{pmatrix}$$

могж. каскад

$$\begin{pmatrix} -\mu & p_\mu \\ 0 & -\mu \end{pmatrix} \begin{pmatrix} -\mu & p_\mu \\ 0 & -\mu \end{pmatrix} = \begin{pmatrix} -2p_\mu^2 \\ 0 \end{pmatrix}$$

$$(e^{-\mu x} (a + b x))'' = (-a \mu e^{-\mu x} + b e^{-\mu x} - b x e^{-\mu x})' \Big|_{x=0}$$

$$= a \mu^2 - 2b \mu \Rightarrow \begin{cases} a \mu^2 - 2b \mu = -2p_\mu^2 \\ -a \mu + b = p_\mu \end{cases}$$

$$\Downarrow a = 0$$

$$b = p_\mu$$

$$f(x) = (10) \begin{pmatrix} e^{-\mu x} p_\mu x e^{-\mu x} \\ 0 e^{-\mu x} \end{pmatrix} \begin{pmatrix} q_\mu \\ \mu \end{pmatrix} = q_\mu e^{-\mu x} + p_\mu^2 x e^{-\mu x}$$

CLT gives normal distribution

T_1, \dots, T_n

$n = 100,000$

$T_{\text{Lokn-1}}$

$\alpha = 0.9$ (90% guarantee)

$T_{1:n} < T_{2:n} < \dots < T_{n:n}$

$T_{1:n}$ gives

Theorem: $(T_\alpha - G(\alpha)) \sqrt{\frac{n}{\alpha(1-\alpha)}} \mathcal{N}(G(\alpha)) \sim N(0, 1), n \rightarrow \infty$

skewness normality

Ex. $n = 100,000$

$\mu = 0.005$

$\alpha = 0.99$

$b = 0.05$

c $p = 0.99$

99% 6 years guarantee

$$P(x) = e^{-\int_0^x \mu(u) du} = e^{-(0.1e^{0.05x} - 0.1)} = 0.01$$

$$\mu(x) = 0.05 \cdot e^{0.05x} \Rightarrow x = 77.025$$

$$G(0.99) = 77.025$$

$$\beta = (x - G(x)) \sqrt{\frac{n}{\alpha(1-\alpha)}} \mathcal{N}(G(x)) = 2.33$$

$$\text{approximate} \Rightarrow x = 77.33$$

X_1, \dots, X_n - i.i.d. $F(x)$

$X_{1:n} \leq \dots \leq X_{n:n}$

$$F_{k:n}(x) = P(X_{k:n} \leq x)$$

$$f_{k:n}(x) = C_n^{k-1} \overbrace{F(x)^{k-1}}^{\stackrel{>}{=}} \overbrace{(1-F(x))^{n-k}}^{\stackrel{>}{=}} = \frac{n!}{(n-k)! k!} F(x)^{k-1} (1-F(x))^{n-k}$$

$$F_{k:n}(x) = P(X_{k:n} \leq x) = \sum_{m=k}^n C_n^m F(x)^m (1-F(x))^{n-m}$$

$$\sum_{m=k}^n C_n^m x^m (1-x)^{n-m} = \int_0^x \frac{n!}{(k-1)!(n-k)!} t^{k-1} (1-t)^{n-k} dt$$

↑ "polys.
patrón"

$$\sum_{m=k}^n \frac{n!}{m!(n-m)!} (m x^{m-1} (1-x)^{n-m} - x^m (n-m) (1-x)^{n-m-1}) =$$

$$\begin{aligned} & \frac{n!}{(k-1)!(n-k)!} t^{k-1} (1-x)^{n-k} + x^k \frac{n!}{k!(n-k-1)!} (1-x)^{n-k-1} \\ & + \frac{n!}{n!(k+1)!(n-k-1)!} (k+1)! x^k (1-x)^{n-k+1} - \dots \end{aligned}$$

l'6a comparar con

$$\int_0^x \frac{n!}{(k-1)!(n-k)!} t^{k-1} (1-t)^{n-k} dt$$

$$f_{k:n}(x) = \frac{n!}{(k-1)!(n-k)!} F(x)^{k-1} (1-F(x))^{n-k} f(x)$$

Regresión $F(x) = \beta x \in [0, 1]$

$$f_{k:n} = \int_0^\infty \frac{n!}{(k-1)!(n-k)!} t^k (1-t)^{n-k} dt = \frac{n!}{(k-1)!(n-k)!} \delta(k+1, n-k+1)$$

$$= \frac{n!}{(k-1)!(n-k)!} \cdot \frac{k! (n-k)!}{(n+1)!} = \frac{k}{n+1} \rightarrow \alpha$$

$$\mathbb{E}[X_{k:n}] = \frac{k(n-k+1)}{(n+1)^2(n+2)} \rightarrow \frac{\alpha(1-\alpha)}{n}$$

$\alpha \xrightarrow{n \rightarrow \infty} k \rightarrow \alpha_n$

U^1, \dots, U^n - i.i.d. $\sim U[0, 1]$

($U_{1:n} \leq U_{2:n} \leq \dots \leq U_{n:n}$)

X_1, \dots, X_{n+1} - i.i.d. $\sim Exp(1)$ $S_k = \sum_{j=1}^k X_j$

$(V_{1:n}, \dots, V_{n:n}) \sim \left(\frac{S_1}{S_{n+1}}, \frac{S_2}{S_{n+1}}, \dots, \frac{S_n}{S_{n+1}} \right)$ distribution
+ joint. repres.

$$P(U_{k:n} \leq x) = P\left(\frac{S_k}{S_{n+1}} \leq x\right) = P(S_k - x S_{n+1} \leq 0) = \phi\left(\frac{n \alpha x}{\sqrt{\alpha n(1-\alpha)^2 + n(1-\alpha)(\alpha^2 + \alpha^2)}}\right)$$

$$E \eta_k = \frac{k(1-x)}{\alpha n} - \frac{(1-x)(n-k)}{(1-\alpha)n} = \frac{\cancel{\alpha n} - \cancel{\alpha n}x - \cancel{\alpha n}x + \cancel{\alpha n}x}{n(\alpha - \alpha x - \alpha^2 - 1)} = \frac{-n x}{\cancel{\alpha} + \cancel{\alpha} + \cancel{\alpha}^2 + \cancel{\alpha}^2}$$

$$\sigma^2(\eta_k) = \alpha n(1-\alpha-\alpha)^2 + \frac{n(1-\alpha)(\alpha+\alpha)^2}{n}$$

$$\sigma^2 = \frac{\alpha(1-\alpha)}{n}$$

$$\approx \phi\left(\frac{\sqrt{n} \cdot x \cdot \sqrt{\alpha(1-\alpha)}}{\sqrt{\alpha n(1-\alpha)^2 + n(1-\alpha)\alpha^2}}\right) = \phi\left(\frac{x \sqrt{n} \sqrt{\alpha(1-\alpha)}}{\sqrt{n} \sqrt{\alpha(1-\alpha)} \sqrt{\alpha + 1 - \alpha}}\right) = \phi(x)$$

A.1.(3)

KP1	20
KP2	20
D2	30
FE	30

CLT give n.p. ch-w

X_1, \dots, X_n

$X_{1:n} < \dots < X_{n:n}$

⊕ give uniform distribution

$$X_\alpha = X_{\lfloor n\alpha \rfloor : n}$$

$$(X_\alpha - G(\alpha)) \sqrt{\frac{n}{\alpha(1-\alpha)}} \text{ of } (G(\alpha)) \sim N(0,1)$$

case $X_1, \dots, X_n \sim U[0,1]$

$$\Rightarrow G(\alpha) = \alpha \quad \text{of } (G(\alpha)) = r$$

$$\Rightarrow (X_\alpha - \alpha) \sqrt{\frac{n}{(1-\alpha)\alpha}} \sim N(0,1)$$

$$(e.g.) \quad u(x) = \mu e^x \quad \mu = 0.005 \quad \beta = 0.05 \quad \Rightarrow \alpha = 0.7$$

$$\begin{aligned} F_x(x) &= y \\ P(Y < x) &= P(F(X) < x) = P(F^{-1}(x)) = \\ &= F(F^{-1}(x)) = x \end{aligned}$$

$F(X_1), \dots, F(X_n) \sim U[0,1]$

$$(F(X_\alpha) - \alpha) \sqrt{\frac{n}{(1-\alpha)\alpha}} \sim N(0,1)$$

$$F(X_\alpha) \stackrel{\text{Taylor}}{=} F(G(\alpha)) + (X_\alpha - G(\alpha)) \text{ of } (G(\alpha)) + o(1)$$

$\Rightarrow \dots$

□

$$\int_0^\infty \frac{\sin \alpha x \sin \beta x}{x} dx = \frac{1}{2} \ln \frac{\alpha + \beta}{|\alpha - \beta|}$$

Trunc approximation

(no convergence) $\nexists \lim f(x)$

$$\int_0^\infty c \underbrace{|f(ax) - f(bx)|}_{\text{numerical error}} dx \quad \text{numerical approximation}$$

$$\Theta (f(0) - f(a)) \ln b/a$$

$$f(0) < \infty$$

$$f(\infty) < \infty$$

$$\textcircled{*} I = \int_0^\infty \frac{1}{x} \int_{bx}^{ax} |f'(z)| dz = \int_0^\infty \frac{|f(ax) - f(bx)|}{x} dx$$

$$\int_0^\infty \frac{e^{-ax} - e^{-bx}}{x} dx = \ln \frac{b}{a}$$

$$z = xy$$

$$\textcircled{a} I = \int_0^\infty \int_B^a |f'(xy)| dy dx$$

$$= \int_B^a dy/y \int_0^\infty |f'(xy)| dx = - \int_a^\infty dy/y (f(\infty) - f(0))$$

$$= (f(0) - f(\infty)) / \ln b/a$$

$$e^{6B(4)} - GBM$$

$$6B(PH) + CTI$$

e

$P(t)$ - уваж. процес. леви а крт. спрем.

$\Gamma(t)$ - Gamma процес

$\Delta \Gamma(t) \sim \text{Gamma}$

$$P(x,t) = \frac{\lambda^{\beta t}}{\Gamma(\beta t)} x^{\beta t - 1} e^{-\frac{x}{\lambda t}}$$

$$\begin{aligned}\varphi_t(\theta) &= E e^{i\theta X_t} = \frac{\alpha \beta^t}{\Gamma(\beta t)} \int_{-\infty}^{\infty} x^{\beta t - 1} e^{i\theta x - \frac{x^2}{2}} dx \\ &= \frac{\alpha \beta^t}{\Gamma(\beta t)} \cdot \frac{1}{(x-i\theta)^{\beta t}} \int_{-\infty}^{\infty} \frac{y}{(x-i\theta)^{\beta t}} e^{-y^2/2} dy = \\ &= \frac{\alpha \beta^t}{(\beta e - i\theta) \Gamma(\beta t)} \int_0^{\infty} y^{\beta t - 1} e^{-y^2/2} dy = \left(1 - \frac{i\theta}{2}\right)^{-\beta t}\end{aligned}$$

Lévy-process subordinator

$$E e^{i\theta X_t} = e^{-t \circ \varphi(\theta)}$$

↓

$$\text{neg. a. way. } \varphi(\theta) = \beta \ln \left(1 - \frac{i\theta}{2}\right)$$

$s < t, X_s, X_t - X_s$

$$E e^{i\theta X_t} = E e^{i\theta X_s}$$

символ употребляется

$$e^{-E(t(\theta))} = E e^{i\theta X_t} = e^{-s \varphi(\theta)} \cdot e^{-(t-s) \varphi(\theta)} =$$

$$= E e^{i\theta X_s} E e^{i\theta t(X_t - X_s)}$$

$$\text{BM: } E e^{itB_t} = E e^{-\frac{\delta^2 t^2}{2}}$$

$$\frac{1}{\sqrt{2\pi t}} \int_{-\infty}^{\infty} e^{i\theta x} e^{-x^2/2t} = e^{-\left(\frac{\delta^2 \theta^2}{2}\right)t}$$

$$B_t = B_s + (B_t - B_s)$$

$$E \left[e^{i\theta B_t} \right] = e^{-\frac{\delta^2 t^2}{2}} \cdot t = e^{-\frac{\delta^2 s^2}{2}} \cdot s \underbrace{e^{-\frac{\delta^2 \theta^2}{2} \cdot (t-s)}}_{= E e^{i\theta(B_t - B_s)}} =$$

$$\psi(\theta) = \int \ln(1 - e^{-\frac{i\theta}{2}}) dt = \int_0^\infty (t e^{i\theta t}) \frac{e^{-\alpha t}}{t} dt$$

$$f(x) = e^{-x} = \int_0^\infty \frac{e^{-\alpha x} - e^{-\alpha x + i\theta t}}{t} dt$$

$$\alpha = -\alpha$$

$$\beta = -\alpha + i\theta$$

$$= \ln \frac{\alpha - i\theta}{\alpha} = \ln \left(1 - \frac{i\theta}{2}\right)$$

$$\psi(\theta) = \int \ln(1 - e^{-\frac{i\theta}{2}}) dt = \int_0^\infty (1 - e^{i\theta x}) \left[\frac{e^{-\alpha x}}{x} \right] dx$$

$\alpha = 2 \Rightarrow BM$

$0 < \alpha \leq 1 \Rightarrow$ Var有限

$\alpha > 1 \Rightarrow$ Var ∞

and $B(tH)$ Var无限

* Levy measure
 $\Rightarrow 0 < \alpha \leq 2 \Rightarrow$ stable ("moments" super nice)
 \Rightarrow tempered stable process

(gamma rubbery property)

$\alpha = 2 \Rightarrow$ var.

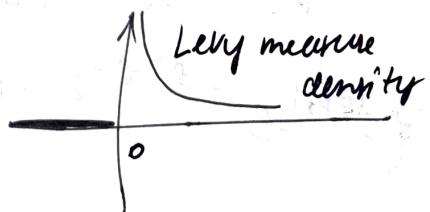
Notation: $\mu \in CP(2..)$

$$S(t) = \sum_{j=1}^{N(t)} Y_j$$

$$E e^{i\theta S(t)} = e^{-t} \int_0^\infty (1 - e^{i\theta x}) f(x) dx$$

$$\int_0^\infty (1 - e^{i\theta x}) \frac{e^{-\alpha x}}{x} dx$$

pack of boxes (countable)
 $\lim_{x \rightarrow 0} \int_0^x \frac{e^{-\alpha x}}{x} dx = \infty$
 no pack. size



$$e^{\delta B(H) + ct}$$

$$\cdot \sum |B_{t_{k+1}} - B_{t_k}|$$

Variation
(Capacities) $|f| = \sup_s \sum_n |f(t_{k+1}) - f(t_k)|$
 \uparrow
 & parcellierung

gut Gamma-Process kapazitiv normen

$$E e^{i\theta [\delta B(H) + CT(H)]}$$

$$\xrightarrow{\text{tower property}} E \left[E \left[e^{i\theta \dots} \middle| \Gamma(t) \right] \right] =$$

$$= E \left[e^{i\theta \left(\frac{\alpha^2 \theta^2}{2} + C\theta \right) \Gamma(t)} \right] =$$

$$= \beta \ln \left(1 - \frac{i(C\theta + \frac{i\theta^2 \alpha^2}{2})}{\alpha} \right) =$$

$$= \beta \ln \left(1 - \frac{i\theta c}{\alpha^{(1)}} + \frac{\theta^2 \alpha^2}{2\alpha} \right) =$$

$$= \beta \ln \left(1 - \frac{i\theta}{\alpha^{(1)}} \right) + \beta \ln \left(1 + \frac{i\theta}{\alpha^{(2)}} \right)$$

$$\alpha^{(2,2)} = \left(\sqrt{\frac{c^2}{4\alpha^1} + \frac{\theta^2}{2\alpha}} \pm \frac{c}{2\alpha} \right)^{-1}$$

$$1 - i\theta \left(\frac{1}{\alpha^{(1)}} - \frac{1}{\alpha^{(2)}} \right) + \frac{\theta^2}{\alpha^{(1)} \alpha^{(2)}} = \frac{c^2}{4\alpha^1} + \frac{\theta^2}{2\alpha} - \frac{c^2}{4\alpha^2}$$

$$Y(t) = \sigma B(T/t) + C P(t) = G_1(t) - G_2(t)$$

$$\beta \int_0^\infty \frac{1-e^{ix}}{x} \frac{e^{-ax}}{x} dx \quad \beta \int_0^\infty \frac{1-e^{-ix}}{x} \frac{e^{-ax}}{x} dx$$

①

$$S(t) = \sum_{j=1}^N Y_j$$

$$E[e^{itS(t)}] = e^{-A} \int_0^\infty (1-e^{i\theta x}) f(x) dx$$

CP omb. можно же сделать
нужные

GP нужно deck сделать
нулевой deck можно просто

$$X \sim (T, \alpha)$$

$$Y \sim (T^*, \alpha^*)$$

$$Z = \min [X, Y]$$

$$S(x) = P(Z > x) = P(X > x) \cdot P(Y > x) =$$

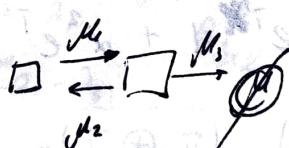
$$= \alpha' e^{T^1 x} \mathbb{1}_n \otimes \alpha^* e^{T^2 x} \mathbb{1}_n =$$

$$= (\alpha' \otimes \alpha^*) (e^{T^1 x} \otimes e^{T^2 x}) \mathbb{1}_{n \times n}$$

$$e^{(T_1 \oplus T_2)x} = Z \sim (\alpha' \otimes \alpha^*; T_1 \oplus T_2)$$

$$T_1 \oplus T_2 = T_1 \otimes I_n + I_n \otimes T_2$$

$$\alpha' = \alpha^* = (1, 0)$$



$$+ \begin{pmatrix} -\mu_1 & \mu_1 \\ \mu_2 & -\mu_2 - \mu_3 \end{pmatrix} \begin{pmatrix} 0 \\ \mu_3 \end{pmatrix} \quad T_2 = \begin{pmatrix} -\mu_1 & \mu_1 \\ \mu_2 & -\mu_2 - \mu_3 \end{pmatrix}$$

$$T_1 = \begin{pmatrix} -\mu_1 & 0 \\ 0 & -\mu_2 - \mu_3 \end{pmatrix}$$

$$A \otimes B = \begin{pmatrix} a_{11}B & \dots & a_{1n}B \\ \vdots & \ddots & \vdots \\ a_{n1}B & \dots & a_{nn}B \end{pmatrix}$$

$$T_1 \oplus T_2 = \begin{pmatrix} -\lambda_1 & \lambda_1 \\ \lambda_3 & -\lambda_2 - \lambda_3 \end{pmatrix} \otimes \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} + \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \otimes \begin{pmatrix} -\mu_1 & \mu_1 \\ \mu_3 - \mu_2 - \mu_3 & \mu_1 \end{pmatrix} =$$

$$= \begin{pmatrix} -\lambda_1 & 0 & \lambda_1 & 0 \\ 0 & -\lambda_1 & 0 & \lambda_1 \\ \lambda_3 & 0 & -\lambda_2 - \lambda_3 & 0 \\ 0 & \lambda_3 & 0 & -\lambda_2 - \lambda_3 \end{pmatrix} + \begin{pmatrix} -\mu_1 & \mu_1 & 0 & 0 \\ \mu_3 & -\mu_2 - \mu_3 & 0 & 0 \\ 0 & 0 & -\mu_1 & \mu_1 \\ 0 & 0 & \mu_3 - \mu_2 - \mu_3 & \mu_1 \end{pmatrix} =$$

$$= \begin{pmatrix} -\lambda_1 - \mu_1 & \mu_1 & 0 & 0 \\ \mu_3 & -\lambda_1 - \mu_2 - \mu_3 & 0 & \lambda_1 \\ \lambda_3 & 0 & -\lambda_2 - \lambda_3 - \mu_1 & \mu_1 \\ 0 & \lambda_3 & \mu_3 & -\mu_2 - \mu_3 \end{pmatrix}$$

$S_2(x)$

$$P(\max\{x, y\} > x) = 1 - (1 - S(x))(1 - S(y)) =$$

$$= 1 - (1 - \alpha^1 e^{T_1 x} \mathbb{1}_n) \otimes (1 - \alpha^2 e^{T_2 x} \mathbb{1}_m) =$$

$$= 1 - \alpha^1 e^{T_1 x} \mathbb{1}_n + \alpha^1 e^{T_1 x} \mathbb{1}_n + \alpha^2 e^{T_2 x} \mathbb{1}_m -$$

$$- (\alpha^1 \otimes \alpha^2) e^{\frac{(T_1 \oplus T_2)x}{\alpha^1 e^{T_1 x} \otimes \alpha^2 e^{T_2 x}}} \mathbb{1}_{mn}$$

$$= \alpha^1 \otimes \alpha^2 e^{\frac{T_1 x}{\alpha^1 e^{T_1 x}}} \mathbb{1}_{mn}$$

$$= \alpha^1 e^{T_1 x} \mathbb{1}_n + \alpha^2 e^{T_2 x} \mathbb{1}_m - \alpha^1 \otimes \alpha^2 e^{\frac{(T_1 \oplus T_2)x}{\alpha^1 e^{T_1 x} + \alpha^2 e^{T_2 x}}} \mathbb{1}_{mn} \quad (\star)$$

$$\alpha^1 e^{T_1 x} = \begin{pmatrix} e^{(T_1 + T_2)x} & (e^{T_1 x} \otimes I_m) - e^{(T_1 + T_2)x} \mathbb{1}_m \\ 0_{n \times n} & 0_{n \times m} \end{pmatrix} \quad \text{and} \quad (I_n \otimes e^{T_2 x}) - e^{\frac{(T_1 + T_2)x}{\alpha^1 e^{T_1 x} + \alpha^2 e^{T_2 x}}} \mathbb{1}_{mn}$$

$$(\alpha^1 \otimes \alpha^2, 0, 0) \cdot e^{\frac{1}{t}x} \cdot \mathbb{1}_n \quad (\star)$$

$\alpha^1 = \alpha^2 = n$

$$\frac{1}{t} = \begin{pmatrix} T_1 \otimes T_2 & T_1 \otimes t^n \\ 0 & T_1 \\ 0 & 0 \\ 0 & T_2 \end{pmatrix}$$

$$\cancel{T_1 \otimes I_n} - \cancel{T_2 \otimes I_n} - I_2 \otimes \cancel{T_2 \mathbb{1}} = I_n \otimes t^n$$

$$= -t_2$$

$$\cancel{I_1 \otimes T_2} - \cancel{T_1 \otimes T_2} - I_1 \otimes I_1$$

$$t = \begin{pmatrix} T & t \\ 0 & 0 \end{pmatrix} \quad t = T \cdot 1$$

$$T = \begin{pmatrix} T_1 \otimes T_2 & \overset{0}{\cancel{-\lambda_1 \mu_3}} & \overset{0}{\cancel{\lambda_1 \mu_3}} & \overset{0}{\cancel{0}} & \overset{0}{\cancel{0}} \\ \cancel{0 \times 4} & \overset{0}{\cancel{\lambda_2 \mu_3}} & \overset{(-\lambda_2 - \lambda_3) \mu_3}{\cancel{(-\lambda_2 - \lambda_3) \mu_3}} & \overset{0}{\cancel{-\lambda_1 \lambda_3}} & \overset{\mu_1 \lambda_3}{\cancel{\mu_1 \lambda_3}} \\ \cancel{0 \times 4} & \overset{0}{\cancel{\lambda_3}} & \overset{\lambda_1}{\cancel{\lambda_1}} & \overset{0}{\cancel{0 \times 2}} & \overset{0}{\cancel{0 \times 2}} \\ 0 \cancel{2 \times 4} & \overset{0}{\cancel{0_{2 \times 2}}} & \overset{0_{2 \times 2}}{\cancel{0_{2 \times 2}}} & \overset{\mu_1 \mu_1}{\cancel{\mu_1 \mu_1}} & \overset{\mu_1 - \mu_2 - \mu_3}{\cancel{\mu_1 - \mu_2 - \mu_3}} \end{pmatrix}$$

$f(x) = \alpha e^{T \otimes T_2} +$

t nach 1 min gehen

$$f(x) = \alpha e^{\frac{1}{t}x} +$$

t nach 2 min gehen

$$\mu_1 = 0,1 \quad \lambda_1 = 0,3$$

$$\mu_2 = 0,3 \quad \lambda_2 = 0,2 \Rightarrow f(x)$$

$$\mu_3 = 0,3 \quad \lambda_3 = 0,1$$

$$Eg = \alpha(-T)^{-1} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

$$(2, 0, 0, 0)$$

$$Eg = \int_0^{\infty} g(x) dx$$

$$\alpha e^{Tx} \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

$$Eg = \int_0^{\infty} x g(x) dx$$

$$\alpha e^{Tx} \cdot t$$

$$t = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0.3 \\ 0 \\ 0.1 \end{pmatrix}$$

$$P(\max[T^1, T^2] > x) = 1 - (1 - s(x))(1 - s(y))$$

$$= 1 - P(T^1 < x) P(T^2 < y) =$$

$$= 1 - (1 - (\alpha^1 e^{Tx} \mathbf{1})) \odot (1 - \alpha^2 e^{Ty} \mathbf{1}) =$$

$$= \alpha^1 e^{Tx} \mathbf{1}_n + \alpha^2 e^{Ty} \mathbf{1}_m - \alpha_1 \odot \alpha_2 e^{T_1 \odot T_2 x} \mathbf{1}_{n \times m}$$

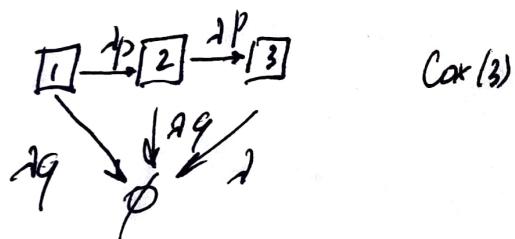
$$e^{Tx} = \begin{pmatrix} e^{T_1 \odot T_2 x} & (e^{T_1 x} \odot I_m - e^{T_1 \odot T_2 x}) \mathbf{1}_m \\ 0 & T_1 \\ 0 & 0 & \ddots & 0 \\ & & & T_2 \end{pmatrix}$$

$$(\alpha' \oplus \alpha^2, 0, 0) \cdot e^{Tx} \cdot \mathbb{1}_{mn} =$$

$$= \alpha_1 e^{T_1 x} \mathbb{1}_n + \alpha_2 e^{T_2 x} \mathbb{1}_m - \alpha_1 \alpha_2 e^{T_1 \oplus T_2 x} \mathbb{1}_{mn}$$

$$\left(\begin{array}{ccc} T_1 \oplus T_2 & T_1 \oplus I_m & T_1 \oplus T_2 \\ 0 & T_1 & 0 \\ 0 & 0 & T_2 \end{array} \right)$$

представление матрицы T



$$\begin{pmatrix} -\lambda & +\lambda p & \lambda q \\ 0 & -\lambda & \lambda p & \lambda q \\ 0 & 0 & -\lambda & \lambda \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

$$\begin{pmatrix} e^{-\lambda x} (a_{11} + b_{12}x + c_{12}x^2) e^{-\lambda x} & \dots \\ 0 & e^{-\lambda x} \\ 0 & 0 & e^{-\lambda x} \end{pmatrix}$$

A. A. (4)

P.H.

Резюме:

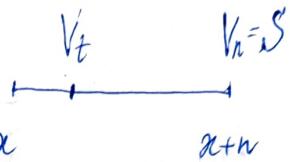
Рез.

Марк.

$$\pi' = \frac{s^T \bar{A}_{x:n}}{\bar{a}_{x:n}}$$

$$\alpha' + \alpha'' s$$

$$\beta' + \beta'' \pi'$$



$$\pi' = \frac{1}{1-f''} \left[\pi + \frac{\alpha' + \alpha'' s}{\bar{a}_{x:n}} + \beta' \right]$$

gross premium

$$e^{iH\delta t} \left[V_t + \underbrace{\left(\pi' - \beta' - \beta'' \pi' \right) \delta t}_{\text{net premium}} \right] = \mu_{x+\delta t} \delta t s + \underbrace{(1 - \mu_{x+\delta t} \delta t)}_u V_{t+\delta t}$$

$$V_{t+\delta t} - \mu_{x+\delta t} \delta t V_t$$

$$V_n = s \quad \leftarrow \text{span. you.}$$

$$\Rightarrow \frac{V_{t+\delta t} - V_t}{\delta t} = (\pi' - \beta' - \beta'' \pi') + \mu_{x+\delta t} V_t - \mu_{x+\delta t} s + \varepsilon V_t$$

$$\frac{dV_t}{dt} = (\pi' - \beta' - \beta'' \pi') + \varepsilon V_t - \mu_{x+\delta t} (s - V_t)$$

$$V_0 = -\alpha' - \alpha'' s$$

$$V_n = \beta'$$

$$\frac{dV_t}{dt} - (\alpha_t \mu_{x+t}) V_t = \pi' - \beta' - \beta'' \pi' - \mu_{x+t} S$$

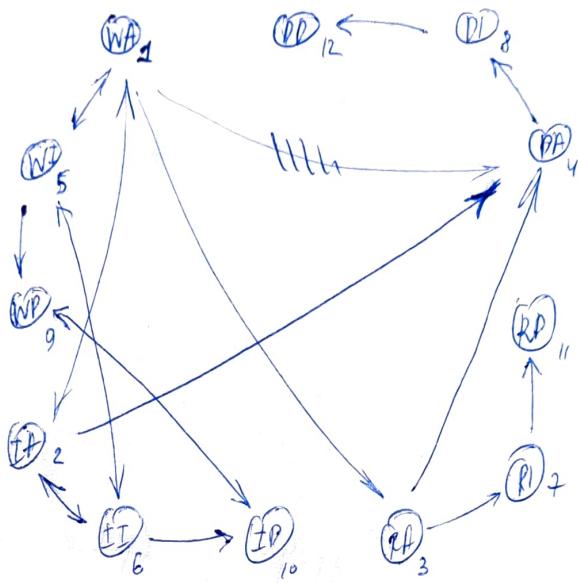
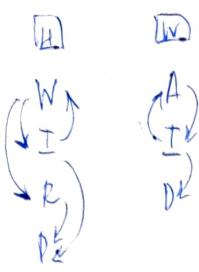
$$e^{\int_t^h (\pi(s) + \mu_{x+s}) ds} \left[-\pi \right] = e^{\int_t^h (\pi(s) + \mu_{x+s}) ds} \left[-\pi \right]$$

$$\int_t^h ds \cdot \frac{d}{dt} \left[e^{\int_t^h (\pi(s) + \mu_{x+s}) ds} V_t \right] = \int_t^h e^{\int_s^h (\pi(u) + \mu_{x+u}) du} - \pi$$

$$V_t = -e^{\int_t^h (\pi(s) + \mu_{x+s}) ds} \left[\int_t^h ds e^{\int_s^h (\pi(u) + \mu_{x+u}) du} (\pi' - \beta' - \beta'' \pi') - \int_t^h \mu_{x+s} e^{\int_s^h (\pi(u) + \mu_{x+u}) du} (-\pi' (1 - \beta'') + \beta' + \mu_{x+s} \cdot S) ds \right. \\ \left. + \int_t^h e^{\int_t^h (\pi(u) + \mu_{x+u}) du} \right]$$

$$V_0 = -\alpha' - \alpha' S \quad (\Rightarrow) \quad \pi' = \frac{1}{1 - \beta''} \left[\underbrace{\alpha' + \alpha' S + \int_0^h e^{-\int_0^s (\pi(u) + \mu_{x+u}) du} \frac{d}{ds} \left(\int_0^s (\pi(u) + \mu_{x+u}) du \right)}_{\text{Simplification}} \right]$$

$$\boxed{\begin{aligned} \int_0^h e^{-\int_0^s (\pi(u) + \mu_{x+u}) du} \mu_{x+u} du &= \int_0^h s e^{-\int_0^s (\pi(u) + \mu_{x+u}) du} du \\ &= \int_0^h e^{-\int_0^s (\pi(u) + \mu_{x+u}) du} (\pi(u) + \mu_{x+u}) du \end{aligned}}$$



$$\begin{pmatrix} Z_1(t) \\ \vdots \\ Z_{11}(t) \end{pmatrix}$$

$$e^{(Z_i(t) + C_i \Delta t) \Delta t} = \sum_j \mu_{ij} \Delta t (f_{ij} + Z_j(t)) -$$

now W

$$- \mu_i \Delta t$$

+ (1 - \sum_j \mu_{ij} \Delta t) Z_i(t + \Delta t)

Kont. neuem

④ reellere gru k=2 gab. an k=1 (gru E)

(10m)

ausarivo

Zam. no norm.

$i = 1, \dots, n$

(Exg normenwerte sind PP)

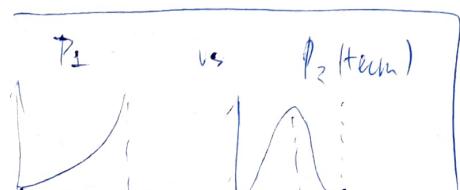
$$E[Z_i(t)] = V_i(t)$$

$$E[Z_i^2(t)] = V_i^2(t)$$

$$(W, A): e^{(V_1(t) + C_1 \Delta t) \Delta t} = \mu_{12} (f + V_2(t)) +$$

$$+ \mu_{13} (d + V_3(t)) + \mu_{15} (f + V_5(t)) +$$

$$+ (1 - \mu_{12} \Delta t - \mu_{13} \Delta t - \mu_{15} \Delta t) V_1(t + \Delta t)$$



$$\frac{dV_1(t)}{dt} = C + f V_1(t) + \mu_{12} (V_2(t) - V_1(t) - f) +$$

$$+ \mu_{13} (V_3(t) - V_1(t) - d) + \mu_{15} (V_5(t) - V_1(t) - f)$$

Umlaufmuster

$$E(M(t) | \mathcal{F}_s) \stackrel{\geq}{\Rightarrow} M(s), \quad s < t$$

\leftarrow super-martingale

$$X_i = \begin{cases} +1 & , p=1/2 \\ -1 & , q=1/2 \end{cases}$$

$$S_n = \sum_{j=1}^n X_j = n(p-q) \quad , \quad p+q=1$$

martingale

$$U_n = (\frac{q}{p})^{S_n}$$

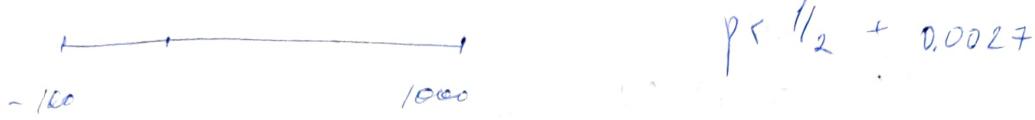
$$E[(\frac{q}{p})^{S_{n+1}} | X_1, \dots, X_n] = E[(\frac{q}{p})^{S_n + X_{n+1}} | X_1, \dots, X_n] =$$

$$= (\frac{q}{p})^{S_n} \cdot E[(\frac{q}{p})^{X_{n+1}}] = (\frac{q}{p})^{S_n} \cdot \underbrace{\left((\frac{q}{p})^1 \cdot p + (\frac{q}{p})^{-1} \cdot q \right)}_{=1} \Rightarrow \text{martingale}$$

$$E(U_\infty) = E(M_\infty) \quad \{ T \leq n \} \in \mathcal{F}_n$$

cup. Spur

$$\tau = \inf \{ h : X_h \in A \}$$



$$\tau = \inf \{ n : S_n = -100 \text{ or } S_n = 1000 \}$$

$$E[(q/p)^{\tau}] = 1$$

$$q^{1000} + (1-q)^{1000} = 1$$

$$q = 1/2 - \epsilon$$

$$p = 1/2 + \epsilon$$

ϵ	0,001	0,003	...
q	0,98	0,6988	...

$$N(t) = B^2(t) - t \quad \leftarrow \text{Martingale}$$

$$M(t) = 3t \cdot B(t) - B^3(t) \quad \leftarrow \text{Martingale}$$

$$E[N(t+\Delta t) | \mathcal{F}_t] = E[B^2(t+\Delta t) - t | \mathcal{F}_t]$$

$$= 3E[(t+\Delta t) \cdot B(t+\Delta t) | \mathcal{F}_t] - E[B^3(t+\Delta t) | \mathcal{F}_t] =$$

$$= 3t E[B(t+\Delta t) | \mathcal{F}_t] + 3\Delta t E[B(t+\Delta t) | \mathcal{F}_t] - E[B^3(t+\Delta t) | \mathcal{F}_t]$$

$$= 3t B_t + 3\Delta t B_t -$$

B.M:

$B(t) \sim \text{ind. stationary increments}$

$$B(t) - B(s) \sim N(0, t-s)$$

$$\stackrel{\sigma}{=} \sqrt{s}$$

Wald Martingal (Begrebt gegen \rightarrow nasseca leben)

$$e^{\lambda B(t) - \frac{\lambda^2 t}{2}}$$

← Martm.

$$e^{\lambda B(t) - \lambda^2 t}$$

← Martm.

$$e^{\lambda B(t) - \frac{\lambda^2 t}{2}}$$

→ DPK
nasseca leben

$$E[B_t - t] = 0$$

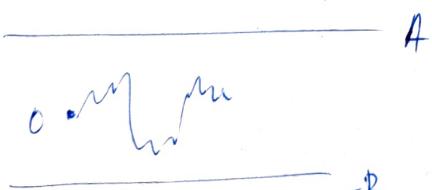
✓

$$E t = A^2$$

$$E e^{\lambda B(t)} = e^{\lambda^2 t}$$

$$E e^{\lambda B(t)} = e^{-t + \frac{\lambda^2 t}{2}}$$

→ nasseca leben



$$\tau = \inf \{t : B(t) = A \text{ or } B(t) = -B\}$$

$$\varphi = P(B(\tau) = A), 1 - \varphi = P(B(\tau) = -B)$$

$$E\tau, E\tau^2$$

$$E[e^{\lambda B(t) - \frac{\lambda^2 t}{2}}] = 1$$

$$2\varphi_2 = s \quad \lambda = \sqrt{2s}$$

$$E[e^{-st}] = \frac{1}{\cosh h(\sqrt{s} \lambda)}$$

$$\frac{1}{2} (e^{\lambda s} + e^{-\lambda s}) E e^{-\frac{\lambda^2 t}{2}} = 1$$

$$\Rightarrow E\tau = A^2 \quad E\tau^2 = BA^2/3$$

$$S(t) = u + \mu t + \sigma B(t) = u - L(t) = k$$

$$e^{SL(t) - \sigma^2 t / 2} \Rightarrow e^{RL(t)} - \text{Montyale}$$

$$\partial S(t) = \partial u + \sigma \partial B(t) = \partial u + \frac{\sigma^2 u}{2} = 0$$

$$1 = \psi(u) + (1 - \psi(u)) e^{R(u-a)}$$

$$\psi(u) = \frac{e^{-2u/\sigma^2} - e^{-2u/\sigma^2 a}}{1 - e^{-2u/\sigma^2 a}}$$

$$\psi(u) = \frac{(1 - e^{R(u-a)}) e^{-Ru}}{(e^{Ru} - e^{R(u-a)}) e^{-Ru}}$$

$$\begin{cases} u=0 \\ \rightarrow \end{cases}$$

$$\text{антипериод } u(0)=1, u(a)=0$$

$$L_{\text{diff}} = y \frac{d}{dx} u + \delta^2/2 \frac{d^2}{dx^2} u \quad | \quad e^{-2u/\sigma^2} \partial$$

$$L_u = \lim_{t \rightarrow 0} \int_0^1 \frac{1}{2\pi t \delta^2} \int_{-\infty}^{\infty} u(x+y) e^{-\frac{(y-u)^2}{2\delta^2 t}} \partial_y u(x)$$

$$L(u(x)) = -\sqrt{\frac{2u}{\sigma^2}} e^{-2u/\sigma^2} x + \delta^2/2 \sqrt{\frac{u}{\sigma^2}} e^{-2u/\sigma^2} x$$

$$L_u = -1 \quad u(0) = u(1) = 0$$

$$L_u = -(1 - \psi(a)), \quad u(a) = u(0) = 0 \Rightarrow E[T] = E[T \cdot 1_{X \leq a}] = E[T \cdot P(X)] / (1 - \psi(a))$$

A. g. (5)

$$E_u(T_a^+ | T_a^+ < t_0^-)$$

$$\chi_u = P_u(T = T_a^+) = \frac{\phi(u)}{\phi(a)} = \frac{1 - \psi(u)}{1 - \psi(a)}$$

$$SH = u + ct - \sum_{j=1}^{Nt} y_j$$

$$\text{L}_u E e^{\theta SH}$$

$$\psi = P_u(T = T_a^+) = \frac{\psi(u) - \chi(a)}{1 - \psi(a)}$$

$$\frac{u}{u + \sqrt{u^2 + 2\psi(u)}}$$

$$E_u \left[e^{-\theta T_a^+} \mathbb{1}_{\{T_a^+ < T_b^+\}} \right] = \frac{W^{(0)}(u)}{W^{(0)}(e)}$$

Scale function

$$LH = \sum y_i - ct$$

$$\text{CGF: } \ln E e^{\theta LH} = \ln$$

$$E e^{i \theta X(H)} = e^{-t \psi(e)}$$

$$= [-ct + \ln E e^{\sum_{j=1}^{Nt} y_j}]$$

$$= -ct + \ln E e^{At(E[e^{\theta y}] - 1)}$$

$$E \left[E e^{\theta \sum_{j=1}^{Nt} y_j} | Nt = 1 \right] =$$

$$= e^{A(E[e^{\theta y}] - 1)}$$

annular debt

$$\hookrightarrow \chi(t) = -ct + A [E[e^{\theta y}] - 1]$$

$$\chi(t) = 0$$

primere ypsilon langschrift
 $\partial F[y](1+\theta) = A(E[e^{\theta y}] - 1)$

$$E[e^{\theta y}] = 1 + E(y)(1+\theta)$$

$$E((E[e^{\theta y}])^n) = \sum_{k=0}^{\infty} e^{-kt} \frac{(At)^k}{k!} E[e^{\theta y}]^n$$

$$= e^{-At} \sum_{k=0}^{\infty} \frac{(At + E[e^{\theta y}])^k}{k!} = e^{At(E[e^{\theta y}] - 1)}$$

\Rightarrow merken wir neu, nochrechnen nötig Dausch

$$e^{\partial X(t) - x(t)} \quad (=) \quad e^{R(X(t))}$$

↑ напр. Решение $x(t) = 0$ к-рекуррентные упр-я для δ_{prob}

$$L(H) = \sum_{j=1}^{N+1} S(H) - ct = u + \underbrace{\xi}_{4+3}$$

$$e^{RL(H)} = 0$$

$$t = \psi(u) E[e^{RL(H)} | T_0 < \infty]$$

$$t = \inf \{ t : L(H) \geq u \}$$

Случайное разрешение

$$t = e^{R_u} \psi(u) E[e^{R_3} | T_0 < \infty]$$

$$\psi(u) = e^{-R_u} \frac{1}{E[e^{R_3} | T_0 < \infty]} \leq e^{-R_u}$$

$$E_u[(T_a^+)^h | T_a^+ < T_0] \xrightarrow{\text{I}} \text{гус. гр.} \quad \text{II} \\ \xrightarrow{\text{II}} \text{scale function} \quad \text{III} \\ \xrightarrow{\text{III}} \text{нагр. напоминание}$$

$$\text{I} \quad \downarrow \text{онеравн. (ненулев.)} \\ T_t d(x) = E_x d(T_t(x))$$

$$T_t \cdot T_s = T_{t+s}$$

$$L = \lim_{t \rightarrow 0} \frac{1}{t} (T_t f - f)$$

↑ онеравн. дифузии упруги с ненулевым нач. в. моментом

$$\text{II} \quad \text{неравн.} \quad X(t) = x + ut + \sigma W(t)$$

$$Ld(x) = \lim_{t \rightarrow 0} \frac{1}{\sqrt{2\pi t}} \int_{-\infty}^{\infty} e^{-\frac{|y-ut|^2}{2\sigma^2 t}} d(x+y) dy - d(x) \quad \text{III}$$

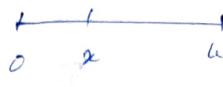
↑ гусеничные параметры со временем \rightarrow экспо \rightarrow $\int_{-\infty}^{\infty} d(x+y) dy$

$$\text{III} \quad u d'(x) + \frac{\sigma^2}{2} d''(x)$$

$$d(x) + \frac{1}{2} d''(x)(y-ut) + \frac{1}{2} d''(x) \left(\frac{\sigma^2 t}{2} \right)^2 + \frac{1}{2} d''(x) [2ut + \frac{\sigma^2 t^2}{2}]$$

$$Lf = 0$$

$$f(0) = 0 \quad f(T=1)$$



$$Lf = 1$$

← сплошные граничные условия

$$f(0) = f(a) = 0$$

$$E [T \mathbb{1}_{\{T_a^+ < T_0^-\}}] \quad \leftarrow \text{среднее сплошные граничные условия } T_a^+ < T_0^-$$

$$\phi_a(x)$$

$$Lf = -\phi_a(x)$$

$$f(a) = \phi(a) = 0$$

где непрерывно дифференцируема

$$Lf(x) = c\phi'(x) + \left[\int_0^\infty \phi(xy) \phi_y(y) dy - \phi(x) \right]$$

$$\text{(II)} \quad E_u \left[e^{-\lambda T_a^+} \mathbb{1}_{\{T_a^+ < T_0^-\}} \right] = \frac{W^{(q)}(u)}{W^{(q)}(a)}$$

$$W^0(x) = \frac{\phi(x)}{\phi'(0)} \quad - \text{1-е приближение}$$

$$Z^{(q)}(u) = 1 + q \int_0^u W^{(q)}(u) du \quad - 2-е приближение,$$

$$\text{Док: } \left[\int_0^\infty e^{-\lambda x} W^{(q)}(x) dx = \frac{1}{\lambda + q} \right]$$

$$E e^{\lambda B(t)} = e^{\frac{\lambda^2}{2} t} \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi t}} e^{\frac{-x^2}{2t}} e^{-\frac{\lambda x^2}{2t}} dx = e^{\frac{\lambda^2}{2} t} \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi t}} e^{-\frac{(x-\lambda t)^2}{2t}} dx = e^{\frac{\lambda^2}{2} t}$$

$$\textcircled{1} \quad x(0) = \frac{\theta^2 \theta^2}{2}$$

$$F^{-1}\left[\frac{1}{\frac{\theta^2 \theta^2}{2} - q}\right] = \sqrt{\frac{2}{q \theta^2}} \sinh \sqrt{\frac{2}{q \theta^2}} x$$

$$\textcircled{2} \quad x(0) = \frac{\theta^2 \theta^2}{2} + \mu \theta$$

$$F^{-1}\left[\frac{1}{\frac{\mu \theta + \theta^2 \theta^2}{2} - q}\right] = \frac{2}{\sqrt{2q \theta^2 + \mu^2}} e^{-\frac{\mu x}{\theta^2}} \sinh \left(\frac{x}{\theta^2} \sqrt{2q \theta^2 + \mu^2} \right)$$

$$T = T_0^+ - \mu T_0^-$$

$$E_u T = ?$$

$$1/2 \int_0^1 f(u) = -1 \quad f = \frac{x(a-u)}{\theta^2} \quad f(0) = f(1) = 0$$

$$\mu f + \frac{\theta^2}{2} f' = -1 \quad Y(H) = f(1) - f(0)$$

$$f(0) = f(1) = 0$$

$$EY(H) = X(H) - f(0)$$

~~see graph!~~

$$X(H) = f(0) + g(H)$$

u

$$a \phi_u(a) \quad \phi_u(a) = \frac{u}{a}$$

$$\text{Lemma } Y(H) = B^2 H - t$$

$$Y(H) = \lambda^2(H) - \frac{\lambda t}{\delta^2}$$

$$Y(t) = u$$

$$(a-u)^2 p(u) + u^2 (1-\varphi(\varepsilon)) = \frac{u}{\delta^2} E(T)$$

$$ET = \frac{1}{\delta^2} \left[(a-u)^2 \frac{u}{a} + u^2 \left(\frac{u-a}{a} \right) \right]$$

$$ET = \frac{u(a-u)}{\delta^2}$$

$$= \frac{1}{\delta^2} u(a-u) \left(\frac{a-u+u}{a} \right) = \frac{u(a-u)}{\delta^2}$$

$$Y(H) \in \mathcal{B}(H)$$

$$T = T_1 \wedge T_0$$

$$ET = \frac{u(a-u)}{\delta^2}$$

$$Y(H) = B^2 H - t$$

$$\delta^2 ET = (a-u)^2 \frac{u}{a} + u^2 \left(1 - \frac{u}{a} \right)$$

$$\frac{df''}{2} = -1$$

$$f(0) = f'(0) = 0$$

$$ET = \phi_u (a-u)^2 \cdot (1 - \varphi_\delta(u)) \delta^2$$

$$ET = \frac{u(a-u)}{\delta^2}$$

e.s

$B^2(t) - t$ \Leftarrow martingale

$$E \left[(H(H) - B(s) + B(s))^2 - (t-s) - s \mid \mathcal{F}_s \right] = B(H) - B(s) = t - s$$

$$\textcircled{1} \quad E_u [e^{-qt} \mathbb{1}_{\{T_a^+ < T_0^-\}}] = \frac{W^{(q)}(u)}{W^{(q)}(a)}$$

$$\frac{W(u)}{W(a)}, \quad q \neq 0 \Rightarrow \frac{u}{a}$$

$$\textcircled{2} \quad - \left(\frac{W(u)}{W(a)} \right) q \quad q=0 \Rightarrow E_u [T_a^+ | T_a^+ < T_0^-] = \frac{x(a^2 - x^2)}{3ax^2}$$

$\left| a^2/2 \sqrt{b} = - \frac{x}{q} \right.$

$f(q) = f'(q) = 0$

$$E_T = \phi_a(u)(a-u)^2 + (1 - \phi_a(u))u^2$$

$$\Rightarrow E_T = \frac{u(a-u)}{a^2}$$

ex (5.2.7)

$$S(H) = a + \mu H + \delta B(H) = u - L(H)$$

$$L(H) = \gamma a t + \delta B(H)$$

↓

$$-\mu H + \frac{\delta^2}{2} \theta^2 = 0$$

$$\theta = \sqrt{\frac{\delta^2}{2}} \theta^2$$

$$\downarrow \text{Нормализация}$$

$$L(H) = a$$

↓

$$E^{\theta} e^{\theta L(T)} = \phi_a(u) e^{(a-u)t} + (-\phi_a(u)) e^{ut}$$

$$\downarrow W^{(u)}(x) = \sqrt{\frac{2}{\pi a^2}} \sin \sqrt{\frac{2}{a^2}} x$$

$$\left\{ \begin{array}{l} \theta^2/2 \theta^4 + \mu \theta^4 = -1 \\ \phi(0) = 0, \phi(a) = 0 \end{array} \right.$$

(5.2.12)

$$E_T = \frac{1}{\mu} [(a-u) \phi_a(u) - u \phi_a(u)]$$

1 new

$$E_u [e^{-q t_0} \mathbb{1}_{\{T_0 < T_0^+\}}] = Z^{(q)}(u) - \frac{W^{(q)}(u)}{W^{(q)}(u)} Z^{(q)}(u)$$

4 (5, 2, 20)

(5.2.23) $W^{(q)}(u) = \frac{2}{\sqrt{2q\delta^2 + \mu^2}} e^{-\frac{\mu u}{\delta^2}} \sinh\left(\frac{u}{\delta^2} \sqrt{2q\delta^2 + \mu^2}\right)$

co cuacony

$$Z^{(q)}(u) = 1 + \int_0^u W^{(q)}(s) ds$$

2. By choice

$$\begin{aligned} & 1 - \phi^{(q)}(x) \quad \text{③ gausse Exp} \\ & \frac{\delta^2}{2} f'' + \mu f' = -\psi^{(q)}(x) \quad \Rightarrow \Sigma = \text{Exp} \\ & f(0) = 0, f(u) = 0 \end{aligned}$$

Ⓐ

$$q + E(\beta T) + \phi(T_+) = u \quad t = \inf\{t+; \phi_t = \varepsilon\}$$

$$= E(\beta)$$

с 2 разумеющим методом
очн.

с 1 не очн.

$E(\beta T) < \infty$ когда момми очн.

$E\left(\frac{\xi_1}{\xi_1 + T}\right) < \infty$ когда ожидаем

(Задача №9: (нужно показать что не бывает)

$$\bullet \tilde{\mu}(dx) = e^x \mu(dx)$$

$$e^{\theta X H} - x(\theta)t$$

$$e^{\theta f(q)}$$

когда θ : $x(\theta) = q \Rightarrow$ равенство $f(q)$

показано - то ищем θ чтобы $x(\theta) = q$

(B, 2, 17)

чтобы наименьшее.

составить B -функции?

3 задача 5.2.10

$$B^2(H) +$$

1) коррелаторы

$$E \left\{ (B(s) + B(t) - B(s))^3 \mid F_s \right\} = B^3(t) - 3tB^2(t)$$

от 2 членов

$$\Delta^3 \Rightarrow 0 \text{ (единственное)}$$

$$- 3s B(s) - 3s(\Delta) - 3(t-s)B(t)$$

A.I. (6)

C Model: $L(t) = \sum y_j - ct$, $y_j \sim \text{Exp}(\mu)$

$$p = \frac{\lambda}{cp} > 1; \text{ npu } q > 0 \quad E[e^{-q\tau(u)} \mathbb{1}_{\tau(u) < \infty}] = \\ = e^{-\Phi(q)u} \left(1 - \frac{\Phi(q)}{\mu} \right)$$

$$\text{je } \Phi(q) = \Theta q = \frac{1}{2} \left[\mu - \frac{\lambda}{c} - \frac{q}{c} + \sqrt{\left(\mu - \frac{\lambda}{c} - \frac{q}{c} \right)^2 + 4 \frac{q\mu}{c}} \right]$$

$$\rightarrow \text{non. reenue } y_{p+1} \left(\frac{\mu}{\mu - \Theta q} - 1 \right) - c\Theta_q = q$$

секундн працеси віда

$$\triangleright k(\theta) = \lambda \left(\frac{\mu}{\mu - \theta} - 1 \right) - ct + np \cdot q - c \text{ npu.}$$

$$k(\theta_q) = q$$

$$\min k(\theta) = \theta_{\min} = \mu - \sqrt{\lambda \mu / c}$$

$$\text{npu } q > 0 \geq k(\theta_{\min}) = \theta_{\min} c / \lambda^{1/2} - ct$$

мопм. Барож $e^{\theta L(t) - k(\theta)t}$

як $\theta = \Phi(q)$; θ_{\min} . є кр. монотон

$$\mu > \lambda/c$$

$$c = \frac{\lambda}{\mu} (1 + \theta)$$

$$\mu - \lambda/c = R$$

$$E_n [T | T < \infty] = \frac{1}{\mu - \lambda/c} \cdot \frac{1}{c} (\lambda/c u + 1)$$

$$\Psi(u) = P(T < \infty) = \frac{1}{cu} e^{-(\mu - \lambda/c)u}$$

$$\text{Var}_n [T | T < \infty] = \left(\frac{1}{\mu - \lambda/c} \right)^2 \frac{1}{c^2} \left(2 \frac{\lambda}{c} \mu u + \frac{\lambda}{c} + u \right)$$

Одн. гаусс.

$$X(t) = \theta(t) + \mu t$$

$$t_s = \inf \{ t : X(t) = s \}$$

$\mu > 0$

$s > 0$

$$F(x, s) = P(t_s < x) = \phi\left(-\frac{s}{\sqrt{x}} + \mu\sqrt{x}\right) + e^{2\mu s} \phi\left(-\frac{s}{\sqrt{x}} - \mu\sqrt{x}\right)$$

$$f(x, s) = \frac{1}{\sqrt{2\pi}} \frac{s}{x^{3/2}} e^{\mu s - \frac{1}{2}(s^2/x + \mu^2 x)}$$

$$\mu = 0 \quad (X(t) = \theta(t))$$

$$t_s \sim \frac{1}{\xi^2}, \quad \xi \sim N(0, \frac{1}{s^2})$$

$$t_s \sim \frac{1}{\xi^2}$$

s —

время же вин.

время на дарев

- ожиданий процесс

$$F = 2\phi\left(-\frac{s}{\sqrt{x}}\right) \quad N(0, 1)$$

$$P\left(\frac{1}{\xi^2} < x\right) = 2\phi\left(\xi \cdot s - \frac{1 \cdot s}{\sqrt{x}}\right) = 2\Phi\left(-\frac{s}{\sqrt{x}}\right)$$

$$Z(t) = e^{\theta X(t)} - \theta X(t) \cdot t \quad \frac{1}{\pi} \ln E\left[e^{\theta X(t)} - \theta \cancel{X(t)} \mu \cdot t\right]$$

$$X(\theta) = q$$

$$E[e^{i\theta t_s}] = e^{-\psi(\theta)s}$$

$$\psi(\theta) = \sqrt{-2i\theta + \mu^2} - \mu$$

$$-\frac{\mu^2}{2} + \mu\sqrt{\mu^2 - 2\theta} = \theta \Rightarrow \mu^2 - 2\mu\sqrt{\mu^2 - 2\theta} = 2\theta = 0$$

$$x = \mu + \sqrt{\mu^2 + 2\theta}$$

$$E[e^{i\theta X(t) + \psi(\theta) t}] = e^{-\psi(\theta) s}$$

? $y(s)$ - cydops. c $\underbrace{\text{норм. расп.}}_{\text{норм.}} (2, 6, 7)$

$$\pi(dx) = \frac{1}{2\pi} \frac{1}{x^{3/2}} e^{-\mu x/2} \mathbb{1}_{\{x > s\}} dx$$

$$E[e^{i\theta X(t)}] = E[e^{i\theta s + (\lambda^2/2 + \mu\lambda)t}] = 1$$

и напр. бывш.

$$\lambda^2/2 + \mu\lambda = -\theta$$

$$\lambda(T_s) = \theta$$

$$E[e^{i\theta s}] = e^{-\theta s}, \text{ где } \frac{\lambda^2}{2} + \mu\lambda = -\theta$$

9

$$\lambda = -\mu + \sqrt{\mu^2 + 2\theta i}$$

$$E[e^{i\theta s}] = e^{-\lambda \psi(\theta)}$$

и в ожидании

$$\psi(\theta) = -\mu + \sqrt{\mu^2 + 2i\theta}$$

$$\int_0^\infty (e^{-ut} - 1) z^{-1-\alpha} dz = -\frac{y}{2} \int_0^\infty e^{-t} z^{-\alpha} dz = -\Gamma(\alpha) u^\alpha$$

Следовательно $\psi(\theta) = i\theta + \frac{\sigma^2 \theta^2}{2} + \int_{-\infty}^{\infty} (1 - e^{i\theta x}) \frac{1}{|x|} \pi(dx)$
 (см. Справ. гл. 1 § 2)

(у супердинамиков $\theta^2 = 0$)

(просто бега)

если $t \Rightarrow$ монотонный

$$\int_{-\infty}^{\infty} \pi(dx) = \infty \quad (\text{if } F(dx))$$

(если иначе \Rightarrow не является Pois.)

$$\int_{-\infty}^{\infty} |x| \pi(dx) < \infty$$

(нечетное & определенное)

$$\int_{-\infty}^{\infty} x^2 \pi(dx) < \infty$$

нечетное

$$E[e^{i\theta \sum_{j=1}^{N(t)} Y_j}] = e^{-\lambda(\int (1 - e^{i\theta x}) \pi(dx)}$$

$$E E[e^{i\theta \sum_{j=1}^{N(t)} Y_j} | N(t)=n] = E[(e^{\theta Y})^n] =$$

$$= \sum_{k=0}^{\infty} e^{-\lambda t} \frac{\lambda t^k}{k!} (E e^{\theta Y})^k = e^{-\lambda t} \sum_{k=0}^{\infty} \frac{(\lambda t + E[e^{\theta Y}])^k}{k!}$$

$$= e^{\lambda t} \left(\int_0^\infty (e^{i\theta x} - 1) F(dx) \right)$$

$$E[e^{i\theta \Psi(t)}] = e^{-t\psi(\theta)} - t\psi'(\theta) \Rightarrow$$

$$\Rightarrow \psi(i\theta) = \int_0^\infty (1 - e^{i\theta x}) F(dx)$$

$$\pi(dx) = \lambda F(dx)$$

\hookrightarrow - Copy. Ross \leftarrow Sehr. unprakt. impd.

$$\pi(dx) = \frac{1}{\sqrt{2\pi}} x^{3/2} e^{-\mu^2 x/2} \mathbb{1}_{x > 0}$$

$$\int_0^\infty (1 - e^{i\theta x}) \pi(dx) = \sqrt{\mu^2 - 2i\theta} - \sqrt{\mu}$$

$$\frac{1}{\sqrt{2\pi}} \int_0^\infty \frac{(1 - e^{i\theta x})}{x^{3/2}} e^{-\mu^2 x/2} dx$$

$$\frac{1}{\sqrt{2\pi}} \int_0^\infty \frac{(e^{-\mu^2 x/2} - e^{i\theta x - \mu^2 x/2})}{x^{3/2}} dx = \sqrt{\mu^2 - 2i\theta} - \sqrt{\mu}$$



5.5.36

Рекуррентные процессы

$X \sim \text{Pois}(2)$

$$E[X_t] = 2t$$

$$\text{Var}[X_t] = 2t$$

сущес. зен.
знач. памп.

$$Y_t = \left\lfloor \frac{N_t}{2} \right\rfloor \quad E(Y_t) \neq \frac{2t}{2}$$

a) меп. кр.
б) DH-исчез

$$\text{Var}(Y_t) \neq \frac{2t}{2}$$

в) непр. зависим

$$\psi(z, t) = E[z^{N(t)}] = \sum_{j=0}^{\infty} z^k P(N(t)=k) = \sum_{j=0}^{\infty} z^k (p(X_1 + \dots + X_n < t))$$

рекуррентные номера

$$s(t) = \sum_{j=1}^n y_j$$

$$N(t) = \sup_n \{ y_1 + \dots + y_n < t \}$$

y_1, y_2, \dots i.i.d. н.п.

лем $Y \sim \text{exp} \Rightarrow \text{Pois}$

$$\psi(z, t) = e^{-st} E[z^{N(t)}] ds =$$

$$= \frac{1}{s} \sum_{j=0}^{\infty} z^j \left(f(s)^k - f^{(k+1)}(s) \right) = \frac{1}{s} (1 - f(s)) \sum_{j=0}^{\infty} z^j f^{(j)}(s)$$

$$\textcircled{1} \int_0^{\infty} e^{-st} \lambda e^{-\lambda t} dt = \frac{1}{s + \lambda}$$

$$f(s) = \frac{(2s)^2}{s+2} \Rightarrow f'(s) = \frac{1 - f(s)}{1 - 2f(s)} \cdot \frac{1}{s}$$

непр. зависим. от Σ эксп.

$$\left(\int_0^\infty e^{-st} E[Z^{N(t)}] dt \right)'|_{z=1} =$$

$$= \int_0^\infty e^{-st} E[N(t)] dt = \hat{m}(s)$$

$$f(s) = \left(\frac{s}{2+s}\right)^2$$

$$\text{Inv. Lap} \left[\frac{A^2}{s^2 + 2s} \right]$$

$$n = \underbrace{\frac{st}{2}}_c - \underbrace{\frac{1}{4}}_c + \underbrace{\frac{e^{-2st}}{4}}_{\text{nonpulsus}}$$

$\left[\frac{14}{3} \right]$: en. new const

$$n = -\frac{1}{3} + \frac{2t}{3} + \frac{e^{-3\sqrt{3}t/2} \cdot \sin(\frac{1}{2}\sqrt{3}t + \phi)}{3\sqrt{3}}$$

Theorie 1

Ph 2

no phys. mean 1

CL vs non. phys. fin. Time min 1

6 amng Variance

1

All. (7)

$$\textcircled{I} \quad N^{(2)}(t) = \frac{1}{2} [NH - \mathbb{1}_{\{NH \text{ odd}\}}]$$

$$E[N^2(t)] = \frac{1}{2} [\lambda t - P(N_{\text{odd}})] = \frac{1}{2} (\lambda t - e^{\lambda t} \frac{1}{2} (e^{\lambda t} - e^{-\lambda t})) = \frac{\lambda t}{2} - \frac{1}{4} + \frac{1}{4}(1 - e^{-2\lambda t})$$

$$P(N_{\text{odd}}) = \sum_{n \text{ odd}} \frac{(\lambda t)^n}{n!} = \frac{1}{2} (e^{\lambda t} - e^{-\lambda t})$$

$$\sum \frac{(\lambda t)^k}{k!} - \sum_{k \text{ even}} \frac{(\lambda t)^k}{k!} + \sum \frac{(\lambda t)^k}{k!}$$

$$\int_0^\infty e^{-st} E[z^{N^{(2)}(t)}] dt = \sum_j z^j [F^{*j}(t) - F^{*(j+1)}(t)]$$

$$F(t) = \exp(\alpha) \cdot \exp(\beta)$$

$$\phi(z, s) = \sum_j z^j [f^{*j}(s) - f^{*(j+1)}(s)] = \frac{1-f(s)}{s} \sum_j z^j f^{*j}(s) = \frac{1-f(s)}{s(1-zf(s))}$$

$$\phi'(z, s) = \dots = \frac{f(s)}{s(1-f(s))} = \frac{s^2/(s+z)^2}{s(1-s^2/(s+z)^2)} = \frac{z^2}{s(s+zs)} = \frac{z^2}{s(s+zs)}$$

$$\int_0^\infty e^{-st} \lambda e^{-\lambda t} dt = \left(\frac{\lambda}{s+\lambda} \right)$$

$$f(s) = \frac{\lambda^2}{(\lambda+s)^2}$$

$$\text{InvLaplace} \left[\frac{\lambda^2}{s^2(s+2\lambda)} \right] = -\frac{1}{\lambda} + \frac{\lambda t}{2} + \frac{1}{4} e^{-2\lambda t}$$

$$(8) \quad N^{(2)}(t) = \int_0^t v(x) dx \quad v(x) dx$$

$$v(x) = \alpha e^{(T+t-\alpha)x} \quad T \sim (\alpha, T)$$

$$\underbrace{\begin{pmatrix} -\lambda & \lambda \\ 0 & -\lambda \end{pmatrix}}_{T=} \underbrace{\begin{pmatrix} 0 \\ \lambda \end{pmatrix}}_{t=} \quad \alpha = (1, 0)$$

$$x \cdot A^2 \cdot e^{-\lambda x} = (1 \ 0) e^{\begin{pmatrix} -\lambda & \lambda \\ 0 & -\lambda \end{pmatrix} x} \begin{pmatrix} 0 \\ \lambda \end{pmatrix}$$

$$N^r x^{n-1} e^{-\lambda x} = (1 \ 0 \dots 0) e^{Tx} \begin{pmatrix} 0 \\ \vdots \\ 1 \end{pmatrix} \quad T = \begin{pmatrix} -\lambda & \lambda & 0 & \dots & 0 \\ -\lambda & -\lambda & \ddots & & \\ 0 & 0 & \ddots & & -\lambda \end{pmatrix} \quad t = \begin{pmatrix} 0 \\ \vdots \\ 1 \end{pmatrix}$$

$$\begin{pmatrix} -1 & 1 \\ 0 & -1 \end{pmatrix} + \begin{pmatrix} 0 \\ 1 \end{pmatrix} (1, 0) = \begin{pmatrix} -1 & 1 \\ 1 & -1 \end{pmatrix}$$

$$e^{\begin{pmatrix} -1 & 1 \\ 1 & -1 \end{pmatrix} \cdot 2t} = \frac{1}{2} \begin{pmatrix} 1+e^{-2it} & 1-e^{-2it} \\ 1-e^{-2it} & 1+e^{-2it} \end{pmatrix}$$

$$EN^{(2)}(t) = \int_0^t \frac{1}{2} (1-e^{-2itx}) dx = \frac{xt}{2} = \frac{1}{4} + \frac{1}{4} e^{-2itx}$$

$$EN^{(3)}(t) = E \lfloor \frac{N(t)}{3} \rfloor = -\frac{1}{3} + \frac{At}{3} + e^{-\frac{At}{2}} \left(\frac{1}{3} \cos\left(\frac{\sqrt{5}}{2}t\right) + \frac{\sqrt{5}}{9} \sin\left(\frac{\sqrt{5}}{2}t\right) \right)$$

$$T+t \alpha \Leftrightarrow \begin{array}{c} \text{circle} \\ \text{with center } T \\ \text{and radius } \alpha \end{array}$$

$$\textcircled{1} \quad E[N^{(2)}(t)] = \underbrace{\frac{1}{4} [E[N(t)]^2]}_{=E_1} - \underbrace{2E[N(t)]}_{=E_2} \underbrace{[1, N(t) \text{ odd}]}_{=E_3} + \underbrace{P(N(t) \text{ odd})}_{=E_3}$$

$$E_1 = At + \frac{A^2 t^2}{2}$$

$$E_3 = \frac{1}{2}(e^{At} - e^{-At}) e^{-At} = \frac{1}{2}(1 - e^{-2At})$$

$$E_2 = e^{-At} \sum_{n \text{ odd}} n \frac{(At)^n}{n!} = At e^{-At} \sum_{n \text{ even}} \frac{(At)^n}{n!} = At e^{-At} \frac{1}{2} (1 + e^{-2At})$$

A.I. (7)

$$1) X(t) = t - \sum_{j=1}^{N(t)}$$

$$E[e^{-\theta X}] = \int_0^\infty E[e^{-\theta x}] F(dx) = \hat{F}[\phi(\theta)]$$

$$\phi(\theta) = \frac{\partial}{\partial x} X(t) - \bar{x}(0)t$$

супр. вангс

$$E[e^{\phi(\theta)x} e^{(x-\theta t)}] = 1$$

$$E[e^{\phi(\theta)x - \theta x}] = 1$$

$$e^{-\phi(\theta)x} E[e^{-\theta x}] = e^{-\phi(\theta)x}$$

$$\text{Следов.: } \frac{1}{t} \ln E[e^{\theta X}]$$

$$= \varphi(\theta)$$

$$= \theta + 2 \int_0^\infty (e^{-\lambda x} - 1) F(dx)$$

$$Ee^{\theta t} \left[Ee^{-\theta \sum_{j=1}^{N(t)} y_j} \mid N(t) = k \right] =$$

$$= e^{\theta t} \sum_{k=0}^{\infty} \frac{(\lambda t)^k}{k!} e^{-\lambda t} (Ee^{-\theta y})^k =$$

$$= e^{\theta t} e^{-\lambda t} \sum_{k=0}^{\infty} \frac{(\lambda t + Ee^{-\theta y})^k}{k!}$$

$$\varphi(\theta) = y$$

$$\theta + 2 \int_0^\infty \mu e^{-\mu x} (e^{-\lambda x} - 1) dx = y$$

$$\theta + \frac{2\mu}{\mu+\lambda} - 2 = y$$

$$\Rightarrow \phi(x)$$

$$\hat{F}(\phi(u)) = \frac{\mu}{\phi(u) + \mu}$$

$$\hat{F}(s) = \int_0^\infty \mu e^{-sx} e^{-\mu x} dx = \frac{\mu}{s+\mu}$$

↑
гип. 2 раз

$$\tau_x = \inf \{t : X(t) = x\}$$

$$(B|X_1=2) \approx \tau_x$$

ψ

$$\text{Var}[B] = \frac{\lambda + \mu}{(\mu - \lambda)^2}$$

$$E[B] = \frac{1}{\mu - \lambda}$$

$$2) U(t) = u + ct - \sum_{j=1}^{N(t)} y_j = S(t)$$

$$(1+\theta) \int_0^t \frac{S(u)}{u} du$$

← азимутальное

направление (нап. ось пак. ядра. со вкл.)

$$S(u)/u = \sum_{j=1}^{N(t)} y_j/u$$

ergodic

$$\mathbb{E} \mu_i < 1$$

$$\dots < \infty$$

$$C = \lambda, \mu, (1+\theta)$$

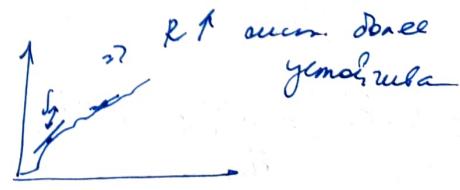
$$\mu_1 = E[X_1]$$

$$(B|X_1=x) \sim T_x$$

безраз.

спекр. омпнс.

$$E[e^{-\theta x}] = e^{-\phi(\theta)x}$$



$$E[e^{Ry}] = 1 + ((1+\theta) E[y])$$

$$E\left[\frac{e^{Ry}}{1 + (1+\theta)y}\right] = 1 \quad y \geq R$$

$$L(H) = \sum_{j=1}^{N(H)} y_j - (1+\theta) \int_0^+ \frac{\sum_{i=1}^{N(H)} y_i}{u} du$$

~~1/2 $\sum_{j=1}^{N(H)} y_j$~~

$$\int_{T_j}^+ du/u = \ln \frac{t}{T_j}$$

$$L(H) = \sum_{j=1}^{N(H)} y_j (1 - (1+\theta) \ln \frac{t}{T_j})$$

$$W_j \sim y_j (1 + (1+\theta) \ln \theta_j)$$

определение оценки симп. интегр.

$$V(H) = \sum_{j=1}^{N(H)} W_j$$

$$E[e^{zH}] = E[e^{zY_j} e^{\ln \theta_j (1+\theta) Y_j}] =$$

$$= E[e^{zY_j} (1 + (1+\theta) Y_j)] =$$

$$= E\left[e^{2Y_j} \int_0^1 e^{(1+\theta) Y_j t} dt\right]$$

$$\{T_j\}_{j=1}^n \sim V(u)$$

$$\text{pair } 5 \xrightarrow{1} \underbrace{t=15}$$

сильно близко расположены - непр. интегр.

$$N(H) = k$$

$$\left(\frac{T^{(1)}}{t}, \frac{T^{(2)}}{t}, \dots, \frac{T^{(k)}}{t}\right)$$

$$\sim (U^{(1)}, U^{(2)}, \dots, U^{(k)})$$

3) Множества генерации: вероятн. независимы

$$S(H) = E\left[\sum_{j=1}^{N(H)} e^{-2H - T_j} y_j \mid N(H)=k\right]$$

запомнил: вероятн. независимы

$$= -2t \left(1 - \frac{T_j}{t}\right)$$

мк. доказать вероятн.

$$E[S(H) \mid N(H)=k] = k \cdot E\left[e^{-2t(1-\frac{1}{t})} y\right] =$$

и дальше

$$= K E\left[\int_0^1 e^{-2t\theta} d\theta\right] = \frac{K}{2t} [1 - e^{-2t}]$$

$$E[K] = H$$

$$E[S(H)] = \frac{H}{2} (1 - e^{-2t})$$

усл. оценка (см. вышеупомянутую проблему)

$$4) S_n = Y_1 + \dots + Y_n$$

$N(t) = \max \{n : S_n \leq t\}$ recurrent point process

$$\hat{N}(t) = \left\lfloor \frac{N(t)}{2} \right\rfloor$$

$$\text{Var } N^{(2)}(t)$$

$$P(NH = k)$$

$$E[Z^{NH}] = \sum_{k=0}^{\infty} z^k \underbrace{F^{*k}(t) - F^{*(k+1)}(t)}_{P(NH=k)}$$

$$\int_0^\infty e^{-st} E[Z^{NH}] dt = \int_0^\infty e^{-st} \sum_{k=0}^{\infty} z^k (-f^{(k)}(-s)) dt$$

$$= \sum_{k=0}^{\infty} z^k \left(\frac{\hat{f}(s)^k}{s} - \frac{\hat{f}'(s)^{k+1}}{s} \right) = \cancel{\frac{\hat{f}(s)}{s}} \frac{(1-\cancel{\hat{f}(s)})}{s} \sum_{k=0}^{\infty} z^k \hat{f}(s)^k =$$

$$= \frac{1 - \hat{f}(1)}{s(1 - z \hat{f}(s))}$$

$$P(\quad, z) \Big|_{z=1} = \frac{\hat{f}(s)}{s(1 - \hat{f}(s))}$$

$$\hat{m}(s) = \int_0^\infty e^{-st} E[NH] dt$$

$$E[N | NH = 1] = \sum_{k=0}^{\infty} k(k+1) \widehat{F^{*k}(t) - F^{*(k+1)}(t)} =$$

$$= \frac{\hat{f}(s)(1-\hat{f}(s))}{s} \sum_{k=0}^{\infty} k(k+1) \hat{f}^{(k+1)} =$$

$$= \frac{\hat{f}(s)(1-\hat{f}(s))}{s} \left(\frac{1}{1-\hat{f}(s)} \right)^2 = \frac{2 \hat{f}(s)}{s(1-\hat{f}(s))^2}$$

$$E[N | NH = 1] - E(N) \sim E(NH)^2 - \text{Var}(NH)$$

AL.(8)

- per.

Bayer vs Predictivity theory (Buhlman)

- битва

Paris (8)

- абмодель

$\theta \in U(1,3)$

Years	Claims
1	1
2	2
3	1

↳ nonloc

$$i=1, \dots, 7 \quad j=1, \dots, h$$

$$\{\theta_i, h X_{ij}\}$$

θ_i - напр. предиктор

(i) cond. ind. for
fixed θ

(ii) $\text{Cor}(X_{ij}, X_{ij'}) \geq 0$

$\text{Cor}(X_{ij}, X_{ij'})$

$$P(B) = \frac{\theta^{\alpha}}{\Gamma(\alpha)} \theta^{\alpha-1} e^{-\beta\theta} \prod_{i=1}^h \frac{\theta^{x_i}}{x_i!} e^{-\theta}$$

$$C \theta^{\alpha + x_i - 1} - \theta (\rho + h)$$

$$\alpha' = \alpha + x_i \quad \beta' = \rho + h$$

$$E[\theta | X] = \frac{\alpha + x_i}{\rho + h}$$

оннатуральное значение

$$\boxed{\frac{\theta(1-\theta)}{n} = \frac{h}{n+\rho}}$$

минимизирующее неравенство

беспр. решен. у. априорного

$$(1-\theta) \frac{x_i}{\rho} + \theta \bar{x}$$

уравн. залогов (напр. Bayes)

Double Poisson \leftrightarrow Poisson - Gamma Model

$$\theta \in G \quad P(N=k|\theta) = \frac{\theta^k}{k!} e^{-\theta}$$

$$\text{Cov}(X_{ij}, X_{ij}) = E[\underset{0}{\overset{n}{\sum}} \text{Var}(\mu_1 + t)]$$

$$Z = \frac{3}{3 + \frac{2}{1/3}} = \frac{1}{3}$$

$E[\theta]$
 $\text{Var}[\theta]$

$$\frac{4}{3} \cdot \frac{1}{3} + 2 \cdot \frac{2}{3} = 1,7778$$

$$1 \cdot \frac{1}{7} + 2 \cdot \frac{6}{7} = 1,8571$$

Bayes

1,7542

1,85084

$$\frac{10}{10 + \frac{2}{1/3}} = \frac{8}{8/3}$$

$$\frac{17}{10} \cdot \frac{5}{8} + 2 \cdot \frac{3}{8} = 1,8125$$

1,80262

1
2
1
3
1
4
0

El Nach

Var(X_{ij}) 16, 1

$$\text{Cor}(X_{ij}, X_{ij}) = \mathbb{E} [\text{cov}(X_{ij}, X_{ij} | \theta_1)] + \\ + \text{cov} [\underbrace{\mathbb{E}(X_i | \theta_1) \cdot \mathbb{E}(X_j | \theta_2)}_{=0}]$$

$$P(N^{th}(t) = k) =$$

$$4.6.16. \int_0^\infty \frac{(\theta t)^k}{k!} e^{-\theta t} \frac{\beta^\alpha}{\Gamma(\alpha)} \theta^{\alpha-1} e^{-\beta t} d\theta =$$

random
 $\theta \geq 0$

for Double Poisson

Var [$N(t)$] $\gg \mathbb{E}[N(t)]$

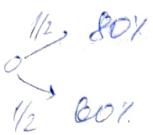
(no same num.
successem)

$$\text{a} \quad \int_0^\infty n^{k+\alpha-1} e^{-n} n^{\alpha-1} du \\ = \Gamma(\alpha+k)$$

$$= \frac{\Gamma(\alpha+k)}{\Gamma(\alpha) \cdot k!} \cancel{\frac{\beta^\alpha t^\alpha}{t^{k+\alpha}}} = \frac{\Gamma(\alpha+k)}{\Gamma(\alpha) \cdot k!} \cdot \left(\frac{\beta}{t}\right)^\alpha ? = \left(\frac{\beta}{\beta+t}\right)^\alpha$$

A. A. (9)

Kelly Betting
Cmabuk m. Leinen



$$80 \cdot 60 = 20\%$$

$$\sum_n 20\% \neq n$$

$$S(1.8^{1/2} \cdot 0.4^{1/2}) = S(1.8^{1/2} \cdot 0.4^{1/2}) = 0.72^{1/2} \rightarrow 0$$

Prüfung Decoupling

$$x_{t+1} = x_t + (k - u_t)$$

$$\left(\begin{array}{l} x_1 = 0 \\ x_0 = s \\ \left(\sum_{t=1}^T x_t \rightarrow \max \right) \\ \boxed{E[\ln x_T] \rightarrow \max} \end{array} \right)$$

$$F_0(x) = \ln x$$

$$F_1(x) = \sup_{0 \leq u \leq x} [\Theta_1 F_0(x+u) + (1-\Theta_1) F_0(x-u)] =$$

$$= \sup_{0 < p \leq 1} [\Theta_1 \ln(x(1+p)) + (1-\Theta_1) \ln(x(1-p))] =$$

$$= \ln x + \sup_{0 < p \leq 1} [\Theta_1 \ln(1+p) + (1-\Theta_1) \ln(1-p)]$$

$$u = p \alpha$$

(Kelly Betting)

$$\Theta = p_{t+1-i}$$

$$\frac{\Theta_1}{1+p} = \frac{1-\Theta_1}{1-p}$$

$$\Theta_1 - \Theta_1 p = 1 - p - \Theta_1 - \Theta_1 p$$

$$\hat{p}_j = 2g_j - 1$$

$$F_2(x) = \ln x + \Theta_1 \Theta_2 (1 + \hat{p}_1 + (1-\Theta_1) \ln(1-\frac{1}{\hat{p}_1}) +$$

$$\Theta_2 \ln(1-\hat{p}_2) + (1-\Theta_2) \ln(1-\frac{1}{\hat{p}_2})$$

$$F_T(x) = \ln x + \sum_{j=1}^T p_j \ln(2p_j) + (1-p_j) \ln(2(1-p_j))$$

$$= \ln x + T \cdot \ln 2 - \sum_{j=1}^T H(p_j)$$

$$\Theta \cdot H(p) = -p \log p - q \log q$$

shannon

Bayara > ~~параметрическим~~ ~~уравнением~~ ~~для~~
 вспомогательного уравнения
 с β и γ можно находит ~~уравнение~~

6.17.

$$1-\beta \quad F = 0$$

$$\beta \quad \max$$

$$\beta = \frac{1}{2} \quad g(x) = \frac{2}{x^3}, \quad x \geq 1$$

и определение негат. и отриц. части.

\Rightarrow Исп. правило вспомогательных функций.

$$\sqrt{\frac{f}{1-f}}$$

$$F(1) = \int_0^\infty \max[y, \beta F(y)] g(y) dy =$$

$$\beta F(1) + \int_0^\infty \max[y - \beta F(y), 0] g(y) dy =$$

$$\beta F(1) + \int_{\beta F(1)}^\infty (y - \beta F(y)) g(y) dy$$

$$(1-\beta) F(1) = \int_{\beta F(1)}^\infty (y - \beta F(y)) g(y) dy$$

$$(1-\beta) F(1) = \int_{\beta F(1)}^\infty \dots = \frac{1}{\sqrt{\beta(1-\beta)}}$$

$$\int_1^\infty \dots \Rightarrow \text{нормальность}$$

$$f(x, t) = \sup_u [c(x, u, t) \cdot ut + e^{-\alpha st} f(x + a(x, u, -1)ut, t + st)]$$

$$f(x, t) = \sup_u [c(x, u, t) \cdot ut + (1 - \alpha)ut] (f(x, t) + \frac{\partial F}{\partial t} \cdot ut + \frac{\partial F}{\partial x} a(x, u, t))$$

$$\sup_u [c(x, u, t) - \alpha t + \frac{\partial F}{\partial t} + \frac{\partial F}{\partial x} \cdot u] = 0$$

(a.s.): $c(x, u, t) = \sqrt{u}$ $a(x, u, t) = \beta x - u$

$$\sup_u [\sqrt{u} - \alpha t + \frac{\partial F}{\partial t} + \frac{\partial F}{\partial x} \cdot (\beta x - u)] = 0$$

$$F(u, t) = \sqrt{u}, \quad f(t) = 0$$

~~$$\frac{1}{2\sqrt{u}} + \frac{\sqrt{t}}{2\sqrt{u}} = 0$$~~

~~$$\sqrt{2\sqrt{u}} = f(t) \quad u = \frac{x}{f^2(t)}$$~~

~~$$F = x/\sqrt{u} \quad u = \frac{\sqrt{x}}{f(t)}$$~~

~~$$\frac{\sqrt{x}}{f(t)} \cdot \left(\frac{1}{2} - \left(x - \frac{1}{2}\beta \right) \cdot f^2 + f'f \right)$$~~

$$Q.1.9. \int_0^T e^{-\alpha t} \sqrt{u(t)} dt \rightarrow \max$$

$$\dot{x}(t) = \beta x(t) - u(t) \quad ; \quad x(T) = 0$$

$$F(u, t) = \sup_u [\sqrt{u} st + e^{-\alpha st} F(x + (\beta x(t) - u(t))st, t + st)]$$

$\frac{\partial x}{\partial u} = (f(x(t)) - u(t)) \cdot st$

$$\dot{F}(x, t) = \sup_u [\sqrt{u} st + ((-\alpha st) F(x, t) + \frac{\partial F}{\partial t} st + \frac{\partial F}{\partial x} f(x) - u) st]$$

$$0 = \sup_u [\sqrt{u} st - \alpha F + \frac{\partial F}{\partial t} + \frac{\partial F}{\partial x} (\beta x - u)] = \sup_u [\sqrt{u} - \alpha x] +$$

$$F(x, t) = \frac{\sqrt{x}}{1 + \frac{1}{2\alpha} f(\beta x - u)}$$

$$f(x) = \frac{1}{\sqrt{x}} + \frac{1}{2\alpha} f'$$

$$f_{uu} = \frac{1}{2x} - \frac{1}{2\alpha} f'' = 0$$

$$u^* = \frac{x}{f}$$

$$\frac{\sqrt{x}}{f} - \alpha \frac{f}{x} = \frac{1}{\sqrt{x}} + \frac{1}{2\alpha} \left(\beta x - \frac{x}{f^2} \right) = 0 \quad (1)$$

$$\frac{1}{f} - \alpha \frac{f}{x} = \frac{1}{x} + \frac{1}{2\alpha} \beta - \frac{1}{2\alpha} \frac{1}{f} = 0$$

$$\frac{1}{f} \left[1 - \alpha \frac{f}{x} + f \frac{1}{x} + \frac{1}{2\alpha} \left(\frac{3}{2} - \frac{1}{f^2} \right) \right] = 0$$

$$\frac{1}{f} \left[1/2 - (\alpha - 1/2)\beta \cdot \frac{f^*}{x} + f \cdot f' \right] = 0$$

$$(f^2)/2 - 2f \cdot f' \cdot 1/2 - f \cdot f' = 0 \quad f^2 = v$$

$$\frac{v'}{2} - (\alpha - \frac{1}{2}\beta)v + \frac{1}{2} = 0$$

$$\frac{v'}{2} = (\alpha - \frac{1}{2}\beta) \cdot v$$

$$\text{d}(t) \int \frac{dv}{v} = 2 \int (\alpha - \frac{1}{2}\beta) dt \cdot v =$$

$$\Rightarrow v = e^{2(\alpha - \frac{1}{2}\beta)t} + C = e^{(2\alpha - \beta)t} + C \quad \text{d}(t) = 0$$

(\Rightarrow)

$$C = \frac{1}{2\alpha - \beta}$$

$$v = \frac{1}{2\alpha - \beta} e^{(2\alpha - \beta)t} + \frac{1}{2\alpha - \beta}$$

$$f(t) = \frac{1 - e^{(2\alpha - \beta)(T-t)}}{2\alpha - \beta}$$

6.1.2 Bang-Bang control

$$u_t = \sum_{t=0}^{h-1} u_t$$

$$u_t = u_t$$

$$0 \leq u_t \leq u_t$$

$$0 \dots 100;$$

$$u_t = a(u_{t+1}) = u_t + \theta(u_t - u_t)$$

$$S = h - t$$

$$F_0(x) = \max_{0 \leq u \leq x} [u + F_{x-u}(x + \theta(x-u))]$$

$$F_0(x) = 0$$

$$F_0(x) = \max_{0 \leq u \leq x} [u + F_0(x + \theta(x-u))] = \max_{0 \leq u \leq x} [u + 0] = x$$

$$F_1 = \max_{0 \leq u \leq x} [u + F_1(x + \theta(x-u))] = \max_{0 \leq u \leq x} [u + x + \theta(x-u)] =$$

$$\text{Bellman's max } x = 0 \vee x \Rightarrow \max_{0 \leq u \leq x} [(1+\theta)x, 2x] = \max_{0 \leq u \leq x} [1+\theta, 2]x = \theta x$$

$$F_3(x) = \max_u [u + 2(\lambda + \theta(x-\omega))] =$$

$$= \max_u [2(\lambda + \theta), 3]x - \rho_3\omega$$

$$F_3(x) = \max [(1+\theta) \cdot p_{s+1} + p_{s-1}]x = p_s x$$

$$p_s = \begin{cases} s, & s \leq s^* \\ (1+\theta)s - s^*, & s \geq s^* \end{cases} \quad \Leftrightarrow (1+\theta)p_{s+1} > 1 - \frac{s}{s^*}$$