

Παράδειγμα

Έστω $N_t \sim PP(\lambda)$ είναι ο αριθμός των σεισμών παγκοσμίως στο $[0, t]$ κ' X_t ο αριθμός των σεισμών στο $[0, t]$ που υπερβαίνουν τα 5 Richter, δείξτε ότι $X_t \sim PP(\lambda p)$ όπου $p = \eta$ πιθανότητα σεισμού ≥ 5 .

$$\begin{aligned} P\{X_t = k\} &= \sum_{n=0}^{\infty} P\{X_t = k | N_t = n\} P\{N_t = n\} = \\ &= \sum_{n=k}^{\infty} P\{X_t = k | N_t = n\} P\{N_t = n\} \end{aligned}$$

όπως: $[X_t | N_t = n] \sim \text{Bin}(n, p)$ έτσι

$$\begin{aligned} P\{X_t = k\} &= \sum_{n=k}^{\infty} \text{Bin}(k | n, p) P_0(n | \lambda t) = \\ &= \sum_{n=k}^{\infty} \binom{n}{k} p^k (1-p)^{n-k} e^{-\lambda t} (\lambda t)^n / n! \\ &= \sum_{n=k}^{\infty} \frac{n!}{k! (n-k)!} p^k (1-p)^{n-k} e^{-\lambda t} (\lambda t)^n / n! \\ &= \frac{e^{-\lambda t} (\lambda t p)^k}{k!} \sum_{n=k}^{\infty} \frac{[\lambda t (1-p)]^{n-k}}{(n-k)!} = \frac{e^{-\lambda t} (\lambda t p)^k}{k!} e^{\lambda t (1-p)} \\ &= e^{-\lambda p t} (\lambda p t)^k / k! = P_0(k | \lambda p t). \end{aligned}$$

Παρατήρηση: $T_i^X \stackrel{iid}{\sim} \text{Exp}(\lambda p)$

Παράδειγμα

As θεωρήσουμε financial market όπου η άφιξη αγορών κόποιων χρεογράφων (risky assets) μοντελοποιείται από $N_t \sim PP(\lambda)$. Το συνολικό αποτέλεσμα N_t οφείλεται στη συγκεντρωτική αγορά χρεογράφων από διαφορετικούς επενδυτές. Εάν υπάρχουν

n ενεργοί επενδυτές $N_t = \sum_{j=1}^n N_t^j$ κ' ο

j -επενδυτής αγοράζει με πιθανότητα p_j , έτσι ώστε

$\sum_{j=1}^n p_j = 1$. Δείξτε ότι $N_t^j \sim PP(\lambda p_j)$ $1 \leq j \leq n$ κ'

$$N_t^i \perp N_t^j, i \neq j$$

$$(i) \quad N_t^j \leq N_t, \forall t \geq 0 \Rightarrow N_0^j \leq N_0 = 0 \Rightarrow P\{N_0^j = 0\} = 1$$

$$\begin{aligned} (ii) \quad P\{N_h^j = 1\} &= \underbrace{P\{N_h = 1\}}_{\lambda h + o(h)} \underbrace{P\{N_h^j = 1 | N_h = 1\}}_{p_j} + \underbrace{P\{N_h \geq 2\}}_{o(h)} P\{N_h^j = 1 | N_h \geq 2\} \\ &= \lambda p_j \cdot h + o(h) \quad (2.1) \end{aligned}$$

$$[N_h^j \geq 2 \Rightarrow N_h \geq 2] \Rightarrow \{N_h^j = 2\} \subseteq \{N_h = 2\} \Leftrightarrow$$

$$\Leftrightarrow P\{N_h^j \geq 2\} \leq P\{N_h \geq 2\} = o(h) \Rightarrow P\{N_h^j \geq 2\} = o(h) \quad (2.2)$$

$$(2.1)(2.2) \Rightarrow P\{N_h^j = 0\} = 1 - \lambda p_j h + o(h)$$

$$\begin{aligned}
 \text{(iii)} \quad P\{N_t^j - N_s^j = k\} &= \sum_{n=k}^{\infty} P\{N_t^j - N_s^j = k \mid N_t - N_s = n\} P\{N_t - N_s = n\} \\
 &= \sum_{n=k}^{\infty} \text{Bin}(k \mid n, p_j) P_0(n \mid \lambda(t-s)) = P_0(n \mid \lambda p_j(t-s)) \\
 &= P\{N_{t-s}^j = n\} \Rightarrow N_t^j - N_s^j \stackrel{d}{=} N_{t-s}^j.
 \end{aligned}$$

$$\begin{aligned}
 \text{(iv)} \quad P\{N_t^j - N_s^j = k \mid N_s^j - N_r^j = \ell\} &= \sum_{n=k}^{\infty} P\{N_t^j - N_s^j = k \mid N_s^j - N_r^j = \ell, N_t - N_s = n\} \\
 &\quad \times P\{N_t - N_s = n \mid N_s^j - N_r^j = \ell\} \\
 &= \sum_{n=k}^{\infty} \text{Bin}(k \mid n, p_j) P_0(n \mid \lambda p_j(t-s)) = \\
 &= P_0(k \mid \lambda p_j(t-s)) = P\{N_t^j - N_s^j = k\}.
 \end{aligned}$$

Παράδειγμα Έστω $N_t \sim PP(\lambda)$ κ' $N_t^j \sim PP(\lambda p_j)$
 τ.ω $N_t = \sum_{j=1}^n N_t^j$ $N_t^i \perp N_t^j$, $i \neq j$, Να βρεθεί ο
 αναμενόμενος χρόνος T παρουσίωσης όλων των τύπων
 αψίδεων.

$$T_1^j = \inf\{t : N_t^j = 1\} = \gamma_1^j$$

$$T = \max_{1 \leq j \leq n} T_1^j$$

$$\begin{aligned}
 P\{T \leq t\} &= P\{T_1^1 \leq t, \dots, T_1^n \leq t\} = \prod_{j=1}^n P\{T_1^j \leq t\} \\
 &= \prod_{j=1}^n (1 - e^{-\lambda p_j t}) \Rightarrow
 \end{aligned}$$

$$E[T] = \int_{\mathbb{R}^+} P\{T > t\} dt = \int_{\mathbb{R}^+} \left\{ 1 - \prod_{j=1}^n (1 - e^{-\lambda_j t}) \right\} dt$$

$$T = \max_{1 \leq j \leq n} T_1^j = \sum_{i=1}^N T_i$$

$$E[T] = E\{E[T|N]\} = E\left\{E\left[\sum_{i=1}^N T_i | N\right]\right\} =$$

$$= E\{N E[T_i]\} = \frac{1}{\lambda} E[N] \Rightarrow \boxed{E[N] = \lambda E[T]}$$

$n=2, \lambda=1$: $E[T] = \int_{\mathbb{R}^+} \left\{ 1 - (1 - e^{-p_1 t})(1 - e^{-p_2 t}) \right\} dt$

$$= 1 + \frac{1}{p(1-p)}, \quad p=p_1, \quad 1-p=p_2.$$

Παράδειγμα: Έστω ότι ταξιδιώτες φτάνουν στο σταθμό τρένου σύμφωνα με την $N_t \sim PP(\lambda)$. Εάν το τρένο αναχωρεί σε χρόνο t , να υπολογιστεί ο ασυμπίεστος χρόνος του χρόνου αναμονής όλων των ταξιδιωτών. $[(Y_1, \dots, Y_n) | N_t = n] \sim \mathcal{U}(0, t, \dots, t)$

$$E\left[\sum_{i=1}^{N_t} (t - Y_i)\right] = E\left\{E\left[\sum_{i=1}^{N_t} (t - Y_i) | N_t\right]\right\} \quad (4.1)$$

$$E\left[\sum_{i=1}^{N_t} (t - Y_i) | N_t = n\right] = nt - E\left[\sum_{i=1}^n Y_i | N_t = n\right] =$$

$$= nt - \mathbb{E} \left[\sum_{i=1}^n \mathcal{U}_{(i)} \right] = nt - \mathbb{E} \left[\sum_{i=1}^n \mathcal{U}_i \right] =$$

$\mathcal{U}_i \stackrel{\text{iid}}{\sim} \mathcal{U}(0, t)$

$$\Rightarrow nt - \frac{nt}{2} = \frac{nt}{2} : (5.1)$$

$$(4.1)(5.1) \Rightarrow \mathbb{E} \left[\sum_{i=1}^{N_t} (t - Y_i) \right] = \mathbb{E} \left[\frac{t}{2} \cdot N_t \right] = \frac{\lambda t^2}{2}$$

Παράδειγμα : Έστω ότι μία αίτηση από $PP(\lambda)$

χαρακτηρίζεται σαν type-I με πιθανότητα $P(s)$

Εάν ευφραίνεται την χρονική στιγμή s κ' type-II

με πιθανότητα $1 - P(s)$. Δείξτε ότι $N_t = N_t^I + N_t^{II}$

με $N_t^I \perp N_t^{II}$. $N_t^I \sim PP(\lambda p)$, $N_t^{II} \sim PP(\lambda(1-p))$ όπου

$$p = \frac{1}{t} \int_0^t P(s) ds$$

$$P \{ N_t^I = n, N_t^{II} = m \} = \sum_{k=0}^{\infty} P \{ N_t^I = n, N_t^{II} = m \mid N_t = k \} P \{ N_t = k \}$$

$$= P \{ N_t^I = n, N_t^{II} = m \mid N_t = n+m \} P \{ N_t = n+m \}$$

$$S = [T_1 \mid N_t = 1] \sim \mathcal{U}(0, t) \Rightarrow$$

$$p = \mathbb{E}[P(S)] = \int_{\mathbb{R}} P(s) f_S(s) ds = \frac{1}{t} \int_0^t P(s) ds \Rightarrow$$

$$P\{N_t^I = n, N_t^{\Pi} = m \mid N_t = m+n\} = \text{Bin}(n+m, p) =$$

\uparrow successes \uparrow failures

$$= \binom{n+m}{n} p^n (1-p)^m$$

Ετσι

$$P\{N_t^I = n, N_t^{\Pi} = m\} = \frac{(n+m)!}{n!m!} p^n (1-p)^m \frac{e^{-\lambda t} (\lambda t)^{n+m}}{(n+m)!}$$

$$= \frac{e^{-\lambda t p} (\lambda t p)^n}{n!} \cdot \frac{e^{-\lambda t (1-p)} [\lambda t (1-p)]^m}{m!}$$

$$= P_0(n \mid \lambda p \cdot t) P_0(m \mid \lambda (1-p) t) = P\{N_t^I = n\} P\{N_t^{\Pi} = m\}$$

Παραδειγμα: Εστω ότι σύστημα υπόκειται σε διαταραχές σύμφωνα με την $N_t \sim PP(\lambda)$. Η i-οστή διαταραχή προκαλεί ζημιά $d_i \stackrel{iid}{\sim} d \sim f(\cdot)$. Η ζημιά όμως λόγω της διαταραχής μειώνεται εκθετικά με την πόροδο του χρόνου δηλ η d_i μετά από χρόνο t γίνεται $d_i e^{-\alpha t}$, $\alpha > 0$. Εάν \mathcal{D}_t είναι η συνολική ζημιά ως κ' χρόνο t τότε

$$\mathcal{D}_t = \sum_{i=1}^{N_t} d_i e^{-\alpha(t-Y_i)}$$

όπου Y_i ο χρόνος αναμονής για την i ομη σειρά Poisson διαταραχή.

$$\mathbb{E}[\mathcal{D}_t \mid N_t = n] = \sum_{i=1}^n \mathbb{E}\{d_i e^{-\alpha(t-Y_i)} \mid N_t = n\} \stackrel{\text{ανεξ.}}{=} \\ = \mathbb{E}[d] e^{-\alpha t} \mathbb{E}\left\{ \sum_{i=1}^n e^{\alpha Y_i} \mid N_t = n \right\}, \underbrace{[Y_i \mid N_t = n]_{i=1}^n}_{\psi_i} \sim \mathcal{U}(0, t)$$

$$\mathbb{E}[D_t | N_t = n] = e^{-\alpha t} \mathbb{E}[d] \mathbb{E}\left\{\sum_{i=1}^n e^{\alpha u_i}\right\} = \mathbb{E}[d] \cdot \frac{n}{t\alpha} (1 - e^{-\alpha t})$$

$$\sum_{i=1}^n \frac{1}{t} \int_0^t e^{\alpha x} dx = \frac{n}{t\alpha} (e^{\alpha t} - 1)$$

$$\mathbb{E}[D_t] = \mathbb{E}\left\{\frac{\mathbb{E}[d]}{t\alpha} N_t (1 - e^{-\alpha t})\right\} = \frac{\lambda \mathbb{E}[d]}{\alpha} (1 - e^{-\alpha t})$$

Παραδειγμα Έαν $\{W_t\}_{t \geq 0}$ η διαδικασία Weiner
 ορίσουμε $\{X_t\}_{t \in [0,1]}$ με $X_t = W_t - tW_1$. Να βρεθεί η
 κατανομή της X_t .

$$P\{W_t - tW_1 \in A\} = \mathbb{E}\left\{P\{W_t - tW_1 \in A | W_1\}\right\} =$$

$$= \int_{\mathbb{R}} P\{W_t - tW_1 \in A | W_1 = u\} N(u|0,1) du =$$

$$= \int_{\mathbb{R}} \left\{ \int_{A+tu} f_{W_t|W_1}(v|u) dv \right\} N(u|0,1) du$$

γνωρίζουμε ότι: $[W_t | W_\tau = y]_{t < \tau} \sim N\left(\frac{t}{\tau} y, \frac{t}{\tau} (\tau - t)\right)$: (7.1)

ΕΤΟΙ $\int_{A+tu} f_{W_t|W_1}(v|u) dv = \int_{A+tu} N(v|tu, t(\tau-t)) dv = \int_A N(v|0, t(\tau-t)) dv$

ΕΤ61 $P\{W_t - tW_1 \in A\} = \int_{\mathbb{R}} \overbrace{N(u/0,1)}^1 du \int_A N(v/0, t(1-t)) dv \Rightarrow$

$\Rightarrow X_t = W_t - tW_1 \sim N(0, t(1-t))$

To covariance ms $\{X_t\}_{t \in [0,1]}$ είναι: για $1 \leq s \leq t \leq 1$

$$\begin{aligned} \text{Cov}(X_s, X_t) &= \text{Cov}(W_s - sW_1, W_t - tW_1) = \\ &= \text{Cov}(W_s, W_t) - t \text{Cov}(W_s, W_1) - s \text{Cov}(W_1, W_t) + st \text{Cov}(W_1, W_1) \\ &= s \wedge t - t \cdot (s \wedge 1) - s(1 \wedge t) + st(1 \wedge 1) = s(1-t) \end{aligned}$$

Ανλ Η X_t είναι μια διαδικασία ορισμένη στο $T=[0,1]$ που είναι Gaussian με μέση τιμή 0 κ' covariance $s(1-t)$, όταν $1 \leq s \leq t \leq 1$.

Παρατηρούμε ότι: $P\{X_1=0\} = P\{X_0=0\} = 1$ κ' ότι

$$X_t = W_t - tW_1 \stackrel{d}{=} [W_t, t \in [0,1] | W_1=0]$$

Πρόκληση: $\text{Cov}[W_s, W_t | W_1=0] =$

$$= \mathbb{E}[W_s W_t | W_1=0] - \underbrace{\mathbb{E}[W_s | W_1=0]}_0 \cdot \underbrace{\mathbb{E}[W_t | W_1=0]}_0 \stackrel{(7.1)}{=} (7.1)$$

$$= \mathbb{E}\left\{ \mathbb{E}[W_s W_t | W_1=0, W_t] \mid W_1=0 \right\} =$$

$$= \mathbb{E} \left\{ \underbrace{\mathbb{E}[W_s W_t | W_t]}_{W_t \underbrace{\mathbb{E}[W_s | W_t]}_{\frac{s}{t} W_t}} \mid W_1 = 0 \right\} = \frac{s}{t} \mathbb{E}[W_t^2 | W_1 = 0] \quad (9.1)$$

$$(7.1) \Rightarrow W_t | W_1 = 0 \sim N(0, t(1-t)) \quad t < 1$$

$$(9.1) \Rightarrow \text{Cov}[W_s, W_t | W_1 = 0] = s(1-t).$$

Παρατήρηση: (i) Δείξτε ότι

$$\text{Cov}(X, Y) = \mathbb{E} \{ \text{Cov}(X, Y | Z) \} + \text{Cov}[\mathbb{E}(X | Z), \mathbb{E}(Y | Z)]$$

$$(ii) \text{ Έστω } Z_t = \sum_{i=1}^{N_t} X_i, \quad X_i \stackrel{i.i.d.}{\sim} X \sim f_X(\cdot)$$

Περίτε το covariance της Z_t .

$$(i) \text{ Cov}(X, Y | Z) = \mathbb{E}[XY | Z] - \mathbb{E}[X | Z] \mathbb{E}[Y | Z] \Rightarrow$$

$$\Rightarrow \mathbb{E} \{ \text{Cov}(X, Y | Z) \} = \mathbb{E}[XY] - \mathbb{E} \{ \mathbb{E}[X | Z] \mathbb{E}[Y | Z] \}$$

$$\begin{aligned} \text{Cov}[\mathbb{E}(X | Z), \mathbb{E}(Y | Z)] &= \mathbb{E} \{ \mathbb{E}(X | Z) \mathbb{E}(Y | Z) \} - \mathbb{E} \{ \mathbb{E}(X | Z) \} \mathbb{E} \{ \mathbb{E}(Y | Z) \} \\ &= \mathbb{E} \{ \mathbb{E}(X | Z) \mathbb{E}(Y | Z) \} - \mathbb{E}[X] \mathbb{E}[Y] \end{aligned}$$

$$(ii) \text{ Cov}(Z_s, Z_t) \stackrel{s \leq t}{=} \mathbb{E} \{ \text{Cov}(Z_s, Z_t | N_t) \} + \text{Cov} \{ \mathbb{E}[Z_s | N_t], \mathbb{E}[Z_t | N_t] \} \quad (9.1)$$

$$\text{Cov}(Z_s, Z_t | N_t) = \mathbb{E} \{ Z_s Z_t | N_t \} - \mathbb{E} \{ Z_s | N_t \} \mathbb{E} \{ Z_t | N_t \} \quad (9.2)$$

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$$\mathbb{E}[Z_s Z_t | N_t = n] = \mathbb{E}\left\{ \sum_{i=1}^{N_s} X_i \cdot \sum_{j=1}^{N_t} X_j \mid N_t = n \right\} = \mathbb{E}\left\{ \sum_{i=1}^N X_i \sum_{j=1}^n X_j \mid N_t = n \right\}$$

οπότε $N = [N_s | N_t = n] \sim \text{Bin}(n, s/t)$, $s < t$. (10.1)

Από την (10.1) γίνεται εμφανές ότι κ' πάλι θα πρέπει να χρησιμοποιήσουμε επαναλαμβανόμενη (υπο-συνθήκη) μέση πμ ή δηλαδή:

$$\mathbb{E}[Z_s Z_t | N_t = n] = \mathbb{E}\left\{ \mathbb{E}[Z_s Z_t | N, N_t = n] \mid N_t = n \right\} \quad (10.2)$$

$$\begin{aligned} \mathbb{E}[Z_s Z_t | N = m, N_t = n] &= \mathbb{E}\left\{ \sum_{i=1}^m X_i \sum_{j=1}^n X_j \right\} = \\ &= \mathbb{E}\left\{ \sum_{i=1}^m X_i \cdot \left(\sum_{i=1}^m X_i + \sum_{j=m+1}^n X_j \right) \right\} = \mathbb{E}\left\{ \left(\sum_{i=1}^m X_i \right)^2 + \sum_{i=1}^m \sum_{j=m+1}^n X_i X_j \right\} \\ &= \mathbb{E}\left\{ \sum_{i=1}^m X_i^2 + 2 \sum_{1 \leq i < j \leq m} X_i X_j + \sum_{i=1}^m \sum_{j=m+1}^n X_i X_j \right\} \\ &= m \mathbb{E}[X^2] + 2 \cdot \binom{m}{2} \mathbb{E}[X]^2 + m(n-m) \mathbb{E}[X]^2 \\ &= m \mathbb{E}[X^2] + m(n-1) \mathbb{E}[X]^2 \end{aligned}$$

$$(10.2) \Rightarrow \mathbb{E}[Z_s Z_t | N_t = n] = \frac{s}{t} \mathbb{E}[X^2] + \frac{n s}{t} (n-1) \mathbb{E}[X]^2 \Rightarrow$$

$$\Rightarrow \mathbb{E}[Z_s Z_t] = \mathbb{E}\left\{ \frac{s}{t} N_t \mathbb{E}[X^2] + \frac{s}{t} N_t (N_t - 1) \mathbb{E}[X]^2 \right\} =$$

$$= \frac{s}{t} \mathbb{E}[N_t] \mathbb{E}[X^2] + \frac{s}{t} \left\{ \mathbb{E}[N_t^2] - \mathbb{E}[N_t] \right\} \mathbb{E}[X]^2$$

$$= \lambda s \mathbb{E}[X^2] + \lambda^2 s t \mathbb{E}[X]^2 : (10.3)$$

$$\begin{aligned} \mathbb{E}[Z_s | N_t = n] &= \mathbb{E} \left\{ \mathbb{E}[Z_s | N, N_t = n] | N_t = n \right\} = \\ &= \mathbb{E} \left\{ N \mathbb{E}[X] | N_t = n \right\} = n \cdot \frac{s}{t} \mathbb{E}[X] \Rightarrow \end{aligned}$$

$$\Rightarrow \mathbb{E}[Z_s | N_t] = \frac{s}{t} \cdot N_t \mathbb{E}[X] \quad , \quad \mathbb{E}[Z_t | N_t] = N_t \mathbb{E}[X]$$

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$$\begin{aligned} \text{Cov} \left\{ \mathbb{E}(Z_s | N_t), \mathbb{E}(Z_t | N_t) \right\} &= \text{Cov} \left\{ \frac{s}{t} N_t \mathbb{E}[X], N_t \mathbb{E}[X] \right\} \\ &= \frac{s}{t} \mathbb{E}[X]^2 \text{Cov}(N_t, N_t) = \frac{s}{t} \mathbb{E}[X]^2 \text{Var}[N_t] = \lambda s \mathbb{E}[X]^2 \end{aligned} \quad (11.1)$$

$$\begin{aligned} \mathbb{E} \left\{ \text{Cov}(Z_s, Z_t | N_t) \right\} &= \mathbb{E} \left\{ \frac{s}{t} N_t \mathbb{E}[X^2] + \frac{s}{t} N_t (N_t - 1) \mathbb{E}[X]^2 - \right. \\ &\quad \left. - \frac{s}{t} N_t \mathbb{E}[X] \cdot N_t \mathbb{E}[X] \right\} = \mathbb{E} \left\{ \frac{s}{t} N_t \text{Var}[X] \right\} = \lambda s \text{Var}[X] \end{aligned} \quad (11.2)$$

$$(9.1)(11.1)(11.2) \Rightarrow \text{Cov}(Z_s, Z_t) \stackrel{s \leq t}{=} \lambda s \text{Var}[X] + \lambda s \mathbb{E}[X]^2 = \lambda s \mathbb{E}[X^2]$$

Από συμμετρία έχουμε $\text{Cov}(Z_s, Z_t) \stackrel{s \geq t}{=} \lambda t \mathbb{E}[X^2] \Rightarrow$

$$\Rightarrow \forall s, t > 0 : \text{Cov}(Z_s, Z_t) = \lambda \cdot s \wedge t \cdot \mathbb{E}[X^2]$$

Παράδειγμα

Δίνεται η σδ $\{K_t\}_{t \in [0, m]}$ κ'

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$$K_t = \inf \left\{ n : \sum_{i=1}^n u_i > t \right\} \quad \text{όπου } u_i \stackrel{\text{i.i.d.}}{\sim} U(0, m). \text{ Να βρεθεί}$$

$$\mu(t) = \mathbb{E}[K_t].$$

$$\begin{aligned} \mu(t) &= \mathbb{E}[K_t] = \mathbb{E}\left\{ \mathbb{E}[K_t | u_1] \right\} = \int_{\mathbb{R}} \mathbb{E}[K_t | u_1 = y] U(y|0, m) dy \\ &= \frac{1}{m} \int_0^m \mathbb{E}[K_t | u_1 = y] dy \end{aligned}$$

$$y > t \Rightarrow \mathbb{E}[K_t | u_1 = y] = 1$$

$$y \leq t \Rightarrow \mathbb{E}[K_t | u_1 = y] = 1 + \mathbb{E}[K_{t-y}] \quad \left. \vphantom{\mathbb{E}[K_t | u_1 = y]} \right\} \Rightarrow$$

$$\Rightarrow \mu(t) = \frac{1}{m} \int_0^m \left\{ 1(y > t) + [1 + \mu(t-y)] 1(y \leq t) \right\} dy =$$

$$= \frac{1}{m} \left\{ m - t + t + \int_0^t \mu(t-y) dy \right\}$$

$$\Leftrightarrow \mu(t) = 1 + \frac{1}{m} \int_0^t \mu(t-y) dy \Rightarrow \mu'(t) = \frac{1}{m} \left[\mu(0) + \int_0^t \mu'(t-y) dy \right]$$

$$\Leftrightarrow \mu'(t) = \frac{1}{m} \left[\mu(0) + \int_0^t \mu'(z) dz \right]$$

$$\Leftrightarrow \mu'(t) = \mu(t)/m \Rightarrow \int_{u=0}^t \frac{d\mu(u)}{\mu(u)} = \int_{u=0}^t \frac{du}{m} \Leftrightarrow$$

$$\Leftrightarrow \mu(t) = \mu(0) e^{t/m} \quad \alpha\lambda\delta \quad \mu(0) = \mathbb{E}[K_0] = \mathbb{E}[1] = 1 \Rightarrow$$

$$\Rightarrow \mu(t) = e^{t/m}$$

□

Παρεμπιπτόν : (i) $\frac{d}{dt} \int_{\alpha(t)}^{\beta(t)} f(t, y) dy = \beta'(t) f(t, \beta(t)) - \alpha'(t) f(t, \alpha(t)) + \int_{\alpha(t)}^{\beta(t)} \frac{\partial}{\partial t} f(t, y) dy$

(ii) Ισχύει πάντα ότι εάν

$$K_t^x = \inf \left\{ n : \sum_{i=1}^n x_i > t \right\}$$

$$N_t^x = \sup \left\{ n : \sum_{i=1}^n x_i \leq t \right\}$$

συνεχώς τη.
 $x_i \stackrel{iid}{\sim} f(\cdot), x_i > 0$

Τότε $P\{K_t^x - N_t^x = 1\} = 1$

Άσκηση Στις σημειώσεις SP1-06.pdf στην άσκηση της σελ. 2] δείξαμε ότι $\text{Cov}(N_s, N_t) = \lambda \cdot s \wedge t$ χρησιμοποιώντας ότι $[N_s | N_t = n] \stackrel{st}{\sim} \text{Bin}(n, s/t)$. (13.1)

Δείξτε ότι $\text{Cov}(N_s, N_t) = \lambda \cdot s \wedge t$ με κανοντας χρήση της (13.1).

$$\begin{aligned} \text{Έστω } s < t \text{ τότε } E[N_s N_t] &= E\{E[N_s N_t | N_s]\} = \\ &= E[N_s E[N_t | N_s]] : (13.2) \end{aligned}$$

$$\begin{aligned} P\{N_t = n | N_s = m\} &= P\{N_t - N_s = n - m | N_s = m\} = P\{N_t - N_s = n - m\} \\ &= P\{N_{t-s} = n - m\} = P_0(n - m | \lambda(t-s)) \end{aligned}$$

$$\begin{aligned} E[N_t | N_s = m] &= \sum_{n=0}^{\infty} n \cdot P\{N_t = n | N_s = m\} = \sum_{n=m}^{\infty} n P\{N_t = n | N_s = m\} \\ &= \sum_{n=m}^{\infty} n \cdot P_0(n - m | \lambda(t-s)) = \sum_{n=m}^{\infty} n \cdot e^{-\lambda(t-s)} \cdot \frac{[\lambda(t-s)]^{n-m}}{(n-m)!} \end{aligned}$$

$$\begin{aligned} &\stackrel{k=n-m}{=} e^{-\lambda(t-s)} \sum_{k=0}^{\infty} (k+m) [\lambda(t-s)]^k / k! = \\ &= e^{-\lambda(t-s)} \left\{ \sum_{k=0}^{\infty} k \cdot [\lambda(t-s)]^k / k! + m \sum_{k=0}^{\infty} [\lambda(t-s)]^k / k! \right\} \end{aligned}$$

$$= e^{-\lambda(t-s)} \left\{ \sum_{k=1}^{\infty} \frac{k}{k!} [\lambda(t-s)]^k + m e^{\lambda(t-s)} \right\}$$

$$= e^{-\lambda(t-s)} \left\{ \lambda(t-s) \underbrace{\sum_{k=1}^{\infty} \frac{[\lambda(t-s)]^{k-1}}{(k-1)!}}_{e^{\lambda(t-s)}} + m e^{\lambda(t-s)} \right\} = \lambda(t-s) + m$$

$$\Rightarrow \boxed{\mathbb{E}[N_t | N_s = m] = \lambda(t-s) + m} \Rightarrow \mathbb{E}[N_t | N_s] \stackrel{s \leq t}{=} \lambda(t-s) + N_s \quad (14.1)$$

$$(13.2) \Rightarrow \mathbb{E}[N_s N_t] = \mathbb{E}[N_s (\lambda(t-s) + N_s)] =$$

$$= \lambda(t-s) \mathbb{E}[N_s] + \mathbb{E}[N_s^2] = \lambda(t-s) \cdot \lambda s + (\lambda s + \lambda^2 s^2)$$

$$= \lambda s + \lambda^2 s t$$

$$\text{ETG1: } \text{Cov}(N_s, N_t) \stackrel{s \leq t}{=} (\lambda s + \lambda^2 s t) - (\lambda s)(\lambda t) = \lambda s$$

απο συμπερασμα: $\text{Cov}(N_s, N_t) = \lambda \cdot s \wedge t, \forall s, t \geq 0$

□

Παρατήρηση: Απο την (14.1) $\Rightarrow \mathbb{E}[N_t - \lambda t | N_s] = N_s - \lambda s$

Αενηση Εστω $X_t \sim PP(\lambda), Y_t \sim PP(s) \quad \kappa'$

$X_t \perp Y_t$ τότε $Z_t = X_t + Y_t \sim PP(\lambda + s)$

$$\begin{aligned} P\{Z_t = n\} &= P\left(\bigcup_{k=0}^n \{X_t = k, Y_t = n-k\}\right) = \sum_{k=0}^n P\{X_t = k, Y_t = n-k\} \\ &= \sum_{k=0}^n P\{X_t = k\} P\{Y_t = n-k\} = \sum_{k=0}^n p_0(k|\lambda t) p_0(n-k|st) \end{aligned}$$

$$\begin{aligned}
&= \sum_{k=0}^n e^{-\lambda t} \frac{(\lambda t)^k}{k!} \cdot e^{-st} \frac{(st)^{n-k}}{(n-k)!} = \frac{e^{-(\lambda+s)t}}{n!} (st)^n \sum_{k=0}^n \binom{n}{k} (\lambda/s)^k \\
&= \frac{e^{-(\lambda+s)t}}{n!} s^n t^n \left(1 + \frac{\lambda}{s}\right)^n = \frac{e^{-(\lambda+s)t}}{n!} \{(\lambda+s)t \}^n \\
&= P_0(n|\lambda+s)
\end{aligned}$$

□

Άσκηση: Δοί (i) Εάν $X_t \sim PP(\lambda)$ κ' $Y_t \sim PP(s)$

$\mu \in X_t \perp\!\!\!\perp Y_t$ τότε $Z_t = X_t + Y_t$ είναι $Z_t \sim PP(\lambda+s)$

(ii) Εάν $X_t \sim CPP(\lambda, f(\cdot))$ κ'

$Y_t \sim CPP(\mu, g(\cdot))$ $\mu \in X_t \perp\!\!\!\perp Y_t$ τότε $Z_t = X_t + Y_t$

είναι $Z_t \sim CPP(\lambda, \frac{\lambda}{\lambda} f(\cdot) + \frac{\mu}{\lambda} g(\cdot))$ όπου $\lambda = \lambda + \mu$.

(iii) Εάν $Y_t \sim CPP(\lambda, \text{Exp}(\mu))$ να βρεθεί

η $M_{Y_t}(s)$ κ' η $\mathbb{E}[Y_t]$

(i) Προηγούμενη άσκηση

(ii) Παραδ. βελ [25]: SP1-05. pdf.

(iii) $Y_t = \sum_{i=1}^{N_t} X_i$; $X_i \stackrel{iid}{\sim} \text{Exp}(\mu) \Leftrightarrow Y_t \sim CPP(\lambda, \text{Exp}(\mu))$

$$M_{Y_t}(s) = \mathbb{E} \left\{ \mathbb{E} \left[e^{s \sum_{i=1}^{N_t} X_i} \mid N_t \right] \right\} = \mathbb{E} \left\{ \mathbb{E} \left[e^{sX_1} \cdots e^{sX_{N_t}} \mid N_t \right] \right\}$$

$$\Rightarrow M_{Y_t}(s) = \mathbb{E} \left\{ M_X(s)^{N_t} \right\} \quad \text{orou} \quad X \sim \text{Exp}(\mu)$$

$$\Leftrightarrow M_{Y_t}(s) = \mathbb{E} \left\{ \exp \left\{ N_t \log(M_X(s)) \right\} \right\} = \left. \begin{aligned} M_X(s) &= \int_{\mathbb{R}^+} e^{sx} (\mu e^{-\mu x} dx) = \frac{\mu}{\mu-s}, \quad s < \mu \end{aligned} \right\} \Rightarrow$$

$$\begin{aligned} M_{Y_t}(s) &= M_{N_t} \left(\log \left(\frac{\mu}{\mu-s} \right) \right) \\ M_{N_t}(s) &= \sum_{k=0}^{\infty} e^{ks} \left\{ e^{-\lambda t} \frac{(\lambda t)^k}{k!} \right\} = e^{-\lambda t} \sum_{k=0}^{\infty} \frac{(\lambda t e^s)^k}{k!} \\ &= e^{-\lambda t} e^{\lambda t e^s} = e^{\lambda t (e^s - 1)} \end{aligned} \quad \Rightarrow$$

$$\Rightarrow \boxed{M_{Y_t}(s) = \exp \left\{ \frac{\lambda t s}{\mu - s} \right\}, \quad s < \mu}$$

$$\begin{aligned} M'_{Y_t}(0) &= \mathbb{E}[Y_t] = \left. \left\{ \frac{\lambda t s}{\mu - s} \right\} \exp \left\{ \frac{\lambda t s}{\mu - s} \right\} \right|_{s=0} = \\ &= \left. \frac{\lambda t \mu}{(\mu - s)^2} M_{Y_t}(s) \right|_{s=0} = \frac{\lambda t}{\mu} M_{Y_t}(0) = \frac{\lambda t}{\mu} \Rightarrow \end{aligned}$$

$$\Rightarrow \boxed{\mathbb{E}[Y_t] = \lambda t / \mu}$$