

Storgruppsövning 14/11-13

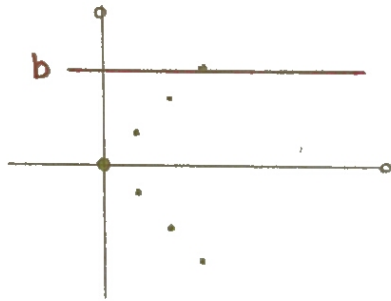
Optional stopping theorem

$(M_n, n \geq 0)$ martingale $(E(M_{n+1} | F_n) = M_n)$
 T stopping time $(T \text{ } \mathbb{N}\text{-valued r.v. } \{T=n\} \text{ is } F_n\text{-measurable})$
 Under "technical conditions" $E(M_T) = E(M_0) (= E(M_n))$

example

Showing that "technical conditions" are needed.
 Consider random walk $X_n = \sum_{i=1}^n Y_i$ where Y_1, Y_2, \dots are IID r.v.'s

$$P(Y_i = -1) = P(Y_i = 1) = 1/2$$



$$T = \min\{n : X_n = b\} \quad b > 0 \text{ is integer}$$

$$b = E(X_T) \neq E(X_0) = 0$$

5.98

Let $X(t)$ be a Poisson process with rate λ .
 Find $P(X(t-d) = k | X(t) = j)$, $0 < t-d < t$

Solution:

$$\begin{aligned} P(X(t-d) = k | X(t) = j) &= \frac{P(X(t-d) = k, X(t) = j)}{P(X(t) = j)} = \\ &= \frac{P(X(t-d) = k, X(t) - X(t-d) = j-k)}{P(X(t) = j)} = \{\text{independent}\} = \\ &= \frac{P(P_0(\lambda(t-d)) = k) P(P_0(\lambda d) = j-k)}{P(P_0(\lambda t) = j)} = \\ &= \frac{(\lambda(t-d))^k}{k!} e^{-\lambda(t-d)} \frac{(\lambda d)^{j-k}}{(j-k)!} e^{-\lambda d} / \frac{(\lambda t)^j}{j!} e^{-\lambda t} = \\ &= \binom{j}{k} \left(1 - \frac{d}{t}\right)^k \left(\frac{d}{t}\right)^{j-k} = P(\text{Bin}(j, 1-d/t) = k) \quad \leftarrow \text{Answer} \end{aligned}$$

5.100

Customers arrive at a bank according to Poisson process with rate $\lambda = 6$ (per/hour). These customers are male with probability $2/3$ and female with probability $1/3$. Suppose we know that 10 men arrived the first two hours. (*) How many women would you expect to have arrived in the first two hours.

Solution:

(*) doesn't matter, they are independent Poisson process.

$\Rightarrow 12/3 = 4$ look at solved problem 5.58

Answer: 4 women.

5.101

Let X_1, \dots, X_n be jointly r.v.'s and let $Y_i = X_i + C_i$ for $i=1, \dots, n$ where C_1, \dots, C_n are constants. Show that Y_1, \dots, Y_n are also jointly normal r.v.'s.

Solution 1:

Def. 1 Z_1, \dots, Z_n jointly normal r.v.'s $\Leftrightarrow \sum_{i=1}^n a_i Z_i$ univariate normal for each choice of constants a_1, \dots, a_n .
 X_1, \dots, X_n are jointly normal $\Rightarrow \sum_{i=1}^n a_i X_i$ normal $\forall a_1, \dots, a_n$.
 $\Rightarrow \sum_{i=1}^n a_i (X_i + C_i)$ normal $\forall a_1, \dots, a_n = \sum_{i=1}^n a_i Y_i$ normal
 $\sum_{i=1}^n a_i X_i + \sum_{i=1}^n a_i C_i \Rightarrow Y_1, \dots, Y_n$ jointly normal \blacksquare

Solution 2:

Def. 2 X_1, \dots, X_n jointly normal if $\Psi_{X_1, \dots, X_n}(w_1, \dots, w_n) = E(e^{i(w_1 X_1 + \dots + w_n X_n)})$
 $= \exp(i w^T \mu_X - \frac{1}{2} w^T K_X w)$
 where $w = \begin{pmatrix} w_1 \\ \vdots \\ w_n \end{pmatrix}$ $\mu_X = \begin{pmatrix} E(X_1) \\ \vdots \\ E(X_n) \end{pmatrix}$ $K_X = \begin{pmatrix} \text{Var}(X_1) & \text{Cov}(X_1, X_2) & \dots & \text{Cov}(X_1, X_n) \\ \text{Var}(X_2) & \text{Cov}(X_2, X_2) & \dots & \text{Cov}(X_2, X_n) \\ \vdots & \vdots & \ddots & \vdots \\ \text{Cov}(X_n, X_1) & \text{Cov}(X_n, X_2) & \dots & \text{Cov}(X_n, X_n) \end{pmatrix}$
 $\Psi_{X_1, \dots, X_n}(w_1, \dots, w_n) = E(e^{i(w_1 X_1 + \dots + w_n X_n)}) =$
 $= E(e^{i(w_1 X_1 + \dots + w_n X_n)} e^{i(w_1 C_1 + \dots + w_n C_n)}) =$
 $= \Psi_{X_1, \dots, X_n}(w_1, \dots, w_n) e^{i(w_1 C_1 + \dots + w_n C_n)} =$
 $= \exp(i w^T \underbrace{(\mu_X + C)}_{\mu_Y} - \frac{1}{2} w^T \underbrace{K_X}_{K_Y} w)$ \blacksquare

5.104

X_1, X_2, \dots are IID r.v.'s, $P(X_i = 3/2) = P(X_i = 1/2) = 1/2$

$M_0 = 1$, $M_n = \prod_{i=1}^n X_i = X_1 \cdot X_2 \cdot \dots \cdot X_n$

Show that $\{M_n, n \geq 0\}$ is a martingale.

Solution:

$E(M_{n+1} | F_n) = M_n$?

F_n = info about the history up to time n . = knowledge of X_1, \dots, X_n = knowledge of M_1, \dots, M_n

$E(M_{n+1} | F_n) = E(\underbrace{X_{n+1}}_{(*)} \underbrace{M_n}_{(+)} | F_n) = \{rule 4\} = M_n E(X_{n+1} | F_n) = \{rule 5\} =$
 $= M_n \underbrace{E(X_{n+1})}_{=1} = M_n$ \blacksquare

(*) - independent of F_n , (+) - F_n -measurable = determined by F_n .

$$E(|M_n|) = E\left(\prod_{i=1}^n |\mathcal{X}_i|\right) = \prod_{i=1}^n \underbrace{E(|\mathcal{X}_i|)}_{=1} = 1$$

5.105

$\mathcal{X}_1, \mathcal{X}_2, \dots$ IID r.v.'s $\begin{cases} P(\mathcal{X}_i = -1) = q = 1-p \\ P(\mathcal{X}_i = 1) = p \end{cases}$

$S_n = \sum_{i=1}^n \mathcal{X}_i$, $Y_n = \left(\frac{q}{p}\right)^{S_n}$, Show Y_n is martingale

(wrt $\mathcal{F}_n = \text{info about } \mathcal{X}_1, \dots, \mathcal{X}_n = \text{info about } Y_1, \dots, Y_n =$
info about S_1, \dots, S_n)

Solution:

$$E(Y_{n+1} | \mathcal{F}_n) = E\left(\underbrace{\left(\frac{q}{p}\right)^{\mathcal{X}_{n+1}}}_{\text{independent of } \mathcal{F}_n} \underbrace{Y_n}_{\text{det. by } \mathcal{F}_n} \mid \mathcal{F}_n\right) = Y_n E\left(\left(\frac{q}{p}\right)^{\mathcal{X}_{n+1}}\right) = Y_n \underbrace{\left(\frac{q}{p} \cdot p + \frac{p}{q} \cdot q\right)}_{q+p=1} =$$

$$= Y_n \quad \blacksquare$$

$$E(\mathcal{X}) = \sum_{\text{all } x} x P(x)$$