1-1) 
$$Z = X_{+}y \rightarrow F_{z}(z) = P(Z(z)) = P(X_{+}y(z)) = \int_{z=-\infty}^{z=-\infty} f_{xy}(x_{y}y) dxdy$$

$$\int_{z=-\infty}^{z=-\infty} \int_{z=-\infty}^{z=-\infty} f_{x}(x) f_{y}(y) dxdy = \int_{z=-\infty}^{z=-\infty} f_{x}(x) \int_{z=-\infty}^{z=-\infty} f_{x}(x) f_{y}(z-x) dx$$

$$\int_{z=-\infty}^{z=-\infty} f_{x}(x) f_{y}(z-x) dx$$

 $f_{z}(z_{-z}) = \frac{1}{2} \int_{z_{-\infty}}^{\infty} f_{x}(x) f_{y}(z_{-x}) dx$ 2-X=t f2(2)- fx(2-t)fx(t) dt

1-2) 
$$J \circ d : \{X - Pois(\lambda)\}$$
  
 $\{Y - Pois(\lambda)\}$ ,  $X \perp Y = \{X + Y - Pois(2\lambda)\}$   
 $P(X = X \mid X + Y = k) = \frac{P(X, X + Y = k)}{P(X + Y = k)} = \frac{P(X = X, Y = k - k)}{P(X + Y = k)} = \frac{P(X = X, Y = k - k)}{P(X + Y = k)} = \frac{P(X = X, Y = k - k)}{P(X + Y = k)} = \frac{P(X = X, Y = k - k)}{P(X + Y = k)}$ 

المنع سال ٢ ما فون السك ( الم) و عدد ( xx) و عدد الم ( الم) و عدد الم المركة عدد الم Auguster our fz(z)= \left\( \frac{1}{2} \right\) \ \ \frac{1}{2} \ \frac{1}{2} \ \ \frac{1}{2} \ \frac{1}{2} \ \frac{1}{2} \ \ \frac{1}{2} ما متفريعان ما سير به المرام مفريعان من الماني هستزولذا لاع، 0< /h= X1+XL < 1 => 0< 26. 24. 24. (الف حال و والام (١٠٠٠ ٤٠٠ / ٢٠١٧ - ١١) مراساب لسر از ما لودن است) وم كسع. fy, y, y, (41, 41, 44) = fx, x1, x1, x1, (x1, x1, x1) x 1/(J(x1, x1, x1))  $\frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} = \begin{vmatrix} \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} & \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} \\ \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} & \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} & \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} \end{vmatrix} = \begin{vmatrix} \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} & \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} & \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} \\ \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} & \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} & \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} \end{vmatrix} = \begin{vmatrix} \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} & \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} & \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} \\ \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} & \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} & \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} \end{vmatrix} = \begin{vmatrix} \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} & \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} & \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} & \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} \\ \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} & \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} & \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} & \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} & \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} \\ \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} & \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} & \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} & \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} & \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} \\ \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} & \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} \\ \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} & \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} & \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} & \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} & \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} \\ \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} & \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} \\ \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} & \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} \\ \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} & \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} \\ \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} & \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J}_{1}} \\ \frac{\partial \mathcal{J}_{1}}{\partial \mathcal{J$ = (y1yr)(y1) fx(x1)fx(/Nx)fx(xx)=y1yr X= -x(x1+xx+xx) = x "y1yre-xy1 ナイノインナーリーンとりはいかしまり افضائه الم ١٤ وولا ٥٥ والاه JY1, Y, Y (41, 4, 4, 4) = 0 Jr 4160 6 44 CO 6 154 fx14 (911 4+) = f fx14.4.4 (4114.4+) = f x giyre-291 dyx = x giyre-291 fy (41) = f fx, xx (4114) dyr = f ly yre 29 dyr = 2 gre - 291

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$$f_{Y_{r}}(y_{r}) = \int_{-\infty}^{+\infty} f_{Y_{r},Y_{r}}(y_{1})y_{r}) dy_{1} = \int_{-\infty}^{+\infty} \lambda^{n}y_{1}^{n}y_{r}e^{-\lambda y_{1}}dy_{1} = -y_{r}(\lambda^{r}y_{1}^{r}+r\lambda y_{1}+r)e^{-\lambda y_{1}^{n}}dy_{1} = -y_{r}(\lambda^{r}y_{1}^{r}+r\lambda y_{1}+r\lambda y_{1}+r)e^{-\lambda y_{1}^{n}}dy_{1} = -y_{r}(\lambda^{r}y_{1}^{r}+r\lambda y_{1}+r\lambda y_{1}+r)e^{-\lambda y_{1}^{n}}dy_{1} = -y_{r}(\lambda^{r}y_{1}^{r}+r\lambda y_{1}+r\lambda y_{1$$

$$f_{X}(x_{1}) = \frac{\lambda^{2}x_{1}^{2}e^{-\lambda x_{1}}}{\lambda^{2}} \qquad f_{X_{1}}(x_{1}) = 1$$

ب. ابتدا سوال را برای حالت  $\mathbb{E}(X)=\mathbb{E}(Y)=0$  حل می کنیم. می داینم:

$$E(Var(Y|X)) = E(E(Y^2|X)) - E(E(Y|X)^2) = E(Y^2) - E(E(Y|X)^2)$$

از طرفی داریم:

$$\rho = \frac{E(XY) - E(X)E(Y)}{\sqrt{Var(XVar(Y))}} = \frac{E(XY)}{\sqrt{E(X^2)E(Y^2)}}$$

پس حکم معادل میشود با:

$$E(Y^2) - E(E(Y|X)^2) \le (1 - \frac{E(XY)^2}{E(X^2)E(Y^2)})E(Y)^2$$
  
 $\iff E(XY)^2 \le E(X^2)E(E(Y|X)^2)$ 

حال دقت كنيد كه:

$$E(XY) = E(E(XY|X)) = E(XE(Y|X))$$

يس حكم معادل است با:

$$E(XE(Y|X))^2 \le E(X^2)E(E(Y|X)^2)$$

که اگ داشته باشیم: W = E(Y|X) حکم معادل می شود با:

$$E(XW)^2 \le E(X^2)E(W^2)$$

که طبق نامساوی کوشی-شوارتز برقرار است. سپس برای  $Z=X-\mathbb{E}(X)$  و  $Z=Y-\mathbb{E}(Y)$  و  $Z=X-\mathbb{E}(X)$  و X و X و X و X و برقراراست، برای X و X هم برقرار است.

Let  $Y = X_1 + X_2 + \cdots + X_n$ , n = 1000.

 $X_i \sim \text{Bernoulli}(p = 0.1)$ 

$$EX_i = p = 0.1$$
  
 $Var(X_i) = p(1 - p) = 0.09$   
 $EY = np = 100$ 

By the CLT:

$$Var(Y) = np(1-p) = 90$$
 as CLT:

$$\frac{Y - EY}{\sqrt{\text{Var}(Y)}} = \frac{Y - 100}{\sqrt{90}}$$
 (can be approximated by  $N(0, 1)$ ). Thus,

$$\sqrt{\operatorname{Var}(Y)}$$
 $P(Y > 125)$ 

$$\sqrt{\text{Var}(X)}$$
 $P(Y > 12)$ 

$$P(Y > 125) = P\left(\frac{Y - 100}{\sqrt{90}} > \frac{125 - 100}{\sqrt{90}}\right)$$

$$P > 125) = P\left(\frac{1}{\sqrt{90}}\right)$$

$$=1-\Phi\left(\frac{25}{\sqrt{90}}\right)$$

 $\approx 0.0042$ 

$$\Phi\left(\frac{25}{25}\right)$$

## مسئلهي ۴. عنوان مسئله

اگر داشته باشیم  $X \sim N(\, {f \cdot}\,, \sigma^{\, \gamma})$  را حساب کنید.

حل.

$$\mathbb{E}(|x^{\mathsf{T}}|) = \int_{-\infty}^{\infty} |x^{\mathsf{T}}| f_X(x) dx = \int_{-\infty}^{\infty} x^{\mathsf{T}} f_X(x) dx - \int_{-\infty}^{\bullet} x^{\mathsf{T}} f_X(x) dx =$$

$$\frac{1}{\sigma\sqrt{\mathbb{Y}\pi}}\left(\int_{\bullet}^{\infty}x^{\mathbb{Y}}e^{-\frac{x^{*}}{\mathbb{Y}\sigma^{*}}}dx-\int_{-\infty}^{\bullet}x^{\mathbb{Y}}e^{-\frac{x^{*}}{\mathbb{Y}\sigma^{*}}}dx\right)$$

بنابراین کافیست مقدار انتگرال زیر را محاسبه کنیم:

$$\int_{a}^{b} x^{7} e^{-\frac{x^{7}}{7\sigma^{7}}} dx$$

با استفاده از انتگرال جزء به جزء داریم:

$$\int_a^b x^{\mathsf{T}} e^{-\frac{x^{\mathsf{T}}}{\mathsf{T}\sigma^{\mathsf{T}}}} dx = \int_a^b x^{\mathsf{T}} . x e^{-\frac{x^{\mathsf{T}}}{\mathsf{T}\sigma^{\mathsf{T}}}} dx = \left[ -\sigma^{\mathsf{T}} x^{\mathsf{T}} e^{-\frac{x^{\mathsf{T}}}{\mathsf{T}\sigma^{\mathsf{T}}}} \right] \Big|_a^b + \mathsf{T}\sigma^{\mathsf{T}} \int_a^b x e^{-\frac{x^{\mathsf{T}}}{\mathsf{T}\sigma^{\mathsf{T}}}} dx$$

$$\int_{a}^{b} x e^{-\frac{x^{*}}{*\sigma^{*}}} dx = \left[ -\sigma^{\mathsf{T}} e^{-\frac{x^{*}}{*\sigma^{*}}} \right]_{a}^{b}$$

$$\implies \int_a^b x^{\mathsf{T}} e^{-\frac{x^{\mathsf{T}}}{\mathsf{T} \sigma^{\mathsf{T}}}} dx = \left[ -\sigma^{\mathsf{T}} x^{\mathsf{T}} e^{-\frac{x^{\mathsf{T}}}{\mathsf{T} \sigma^{\mathsf{T}}}} \right] \Big|_a^b + \mathsf{T} \sigma^{\mathsf{T}} \left[ -\sigma^{\mathsf{T}} e^{-\frac{x^{\mathsf{T}}}{\mathsf{T} \sigma^{\mathsf{T}}}} \right] \Big|_a^b$$

$$\implies \begin{cases} \int_{-\infty}^{\infty} x^{\intercal} e^{-\frac{x^{\intercal}}{10^{\intercal}}} dx = \Upsilon \sigma^{\intercal} \\ \int_{-\infty}^{\infty} x^{\intercal} e^{-\frac{x^{\intercal}}{10^{\intercal}}} dx = -\Upsilon \sigma^{\intercal} \end{cases}$$

$$\implies \mathbb{E}(|x^{\mathsf{T}}|) = \mathsf{Y}\sqrt{\frac{\mathsf{Y}}{\pi}}\sigma^{\mathsf{T}}$$

(a)	(1 pt) hypoth	Constructions.	et a	sample	statistic	which	has	zero	mean	under	the	null

Solution: 
$$S = \bar{X} - .5\bar{Y}$$
.

(b) (2 pts) What is its variance under the null?

Solution: 
$$var(S) = 16/100 + 9/100 = .25$$
.

(c) (2 pts) Construct a rejection region at 95% significance level.

Solution: 
$$P(|S| \ge \xi; H_0) = .05$$
. Under the null,  $S \sim N(0, .25)$ . So  $\xi/.5 = 1.96$  and so  $\xi = .98$ .

(d) (1 pt) Would you accept or reject the null hypothesis at the 95% significance level?

Solution: |S| = |48 - 50| = 3. Clearly in the rejection region. reject

الف

$$f_{X_t}(x;\theta) = \frac{1}{\sqrt{2\pi\theta}}e^{-\frac{x^2}{2\theta}}$$

تابع درست نمایی را بدست میآوریم:

u

$$\begin{split} E[X_i^2] &= \int_{-\infty}^{\infty} x^2 f_{X_i}(x) dx \\ &= \int_{-\infty}^{\infty} x^2 \frac{1}{\sqrt{2\pi\theta}} e^{-\frac{\pi^2}{2\theta}} dx \\ &= -\frac{\theta}{\sqrt{2\pi\theta}} \int_{-\infty}^{\infty} x d(e^{-\frac{\pi^2}{2\theta}}) \\ &= -\frac{\theta}{\sqrt{2\pi\theta}} \underbrace{\left(x e^{-\frac{\pi^2}{2\theta}}\right)_{-\infty}^{\infty}}_{=0} - \int_{-\infty}^{\infty} e^{-\frac{\pi^2}{2\theta}} dx) \\ &= \theta \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi\theta}} e^{-\frac{\pi^2}{2\theta}} dx \\ \Longrightarrow E[X_i^2] &= \theta \end{split}$$

$$\begin{split} B(\hat{\Theta}) &= E[\hat{\Theta}] - \theta \\ &= E[\frac{1}{n} \sum_{i=1}^{n} X_{i}^{2}] - \theta \\ &= \frac{1}{n} \sum_{i=1}^{n} E[X_{i}^{2}] - \theta \\ \Longrightarrow B(\hat{\Theta}) &= 0 \end{split}$$

$$\begin{split} E[X_i^4] &= \int_{-\infty}^{\infty} x^4 f_{X_i}(x) dx \\ &= \int_{-\infty}^{\infty} x^4 \frac{1}{\sqrt{2\pi\theta}} e^{-\frac{x^2}{2\theta}} dx \\ &= -\frac{\theta}{\sqrt{2\pi\theta}} \int_{-\infty}^{\infty} x^3 d(e^{-\frac{x^2}{2\theta}}) \\ &= -\frac{\theta}{\sqrt{2\pi\theta}} \underbrace{\left(x^3 e^{-\frac{x^2}{2\theta}}\right|_{-\infty}^{\infty} - 3 \int_{-\infty}^{\infty} x^2 e^{-\frac{x^2}{2\theta}} dx\right)}_{=0} \\ &= 3\theta \int_{-\infty}^{\infty} x^2 \frac{1}{\sqrt{2\pi\theta}} e^{-\frac{x^2}{2\theta}} dx \\ &= 3\theta E[X_i^2] \\ \Longrightarrow E[X_i^4] = 3\theta^2 \end{split}$$

$$Var(X_i^2) = E[X_i^4] - (E[X_i^2])^2 = 2\theta^2$$
 ... با توجه به مستقل بودن  $X_i$ ها از یکادیگر مستقل اند. 
$$Var(\hat{\Theta}) = Var(\frac{1}{n}\sum_{i=1}^n X_i^2)$$
 
$$= \frac{1}{n^2}\sum_{i=1}^n Var(X_i^2)$$
  $\Longrightarrow Var(\hat{\Theta}) = \frac{2\theta^2}{n}$ 

$$MSE(\hat{\Theta}) = Var(\hat{\Theta}) + B(\hat{\Theta})^2 = \frac{2\theta^2}{n}$$
  
 $\Longrightarrow \lim_{n\to\infty} MSE(\hat{\Theta}) = 0$ 

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