Problem 1

(a)

$$X_1, \dots, X_n \sim U(0, \theta)$$

$$f_X(x) = \frac{1}{\theta} \quad 0 < \theta < \infty$$

$$X_{(1)}, \dots, X_{(n)} \text{ are order statistics}$$

$$Y_i = \frac{X_i}{X_{i+1}} \text{ for } i = 1, \dots, n-1$$

$$Y_n = X_{(n)}$$

$$Y_1 = \frac{X_{(1)}}{X_{(2)}} \quad Y_{n-1} = \frac{X_{(n-1)}}{X_{(n)}}$$

$$Y_n = X_{(n)}$$

$$X_{(n)} = Y_n \quad X_{(n-1)} = Y_n Y_{n-1} \quad \cdots \quad X_1 = Y_1 Y_2 \cdots Y_n$$

$$X_{(i)} = \prod_{j=i}^n Y_j$$

$$J = \begin{bmatrix} \frac{dx_1}{dy_1} & \dots & \frac{dx_1}{dy_n} \\ \frac{dx_2}{dy_1} & \dots & \dots \\ \dots & \dots & \dots \\ \frac{dx_n}{dy_1} & \dots & \frac{dx_n}{dy_n} \end{bmatrix}$$

$$= \begin{bmatrix} y_2 \cdots y_n & \dots & y_1 \cdots y_{n-1} \\ 0 & y_3 \cdots y_n & y_2 \cdots y_{n-1} \\ \dots & \dots & \dots \\ 0 & \dots & 1 \end{bmatrix}$$
even triangular matrix thus its determinant in the pro-

J is an upper triangular matrix thus its determinant is the product of its main diagonal $J = |y_2 y_3^2 \dots y_n^{n-1}| = y_2 y_3^2 \dots y_n^{n-1}$

(b)

$$f_{Y_1,\dots,Y_n}(y_1,\dots,y_n) = f_{X_{(1)},\dots,X_{(n)}}(\prod_{j=1}^n (y_j,\dots,y_n))|J|$$
$$= \frac{n!}{\theta^n}(y_2y_3^2\dots y_n^{n-1})$$

This can be factored into functions of y_i for i = 1, ..., n

Thus the Ys are mutually independent

(c)

Since there is no
$$y_1$$
 term:
$$f_{Y_1}(y_1) = 1 \quad 0 < y_1 < 1$$

$$f_{Y_i}(y_i) = iy_i^{i-1} \quad 0 < y_i < 1$$

$$f_{Y_n}(y_n) = n\frac{y_n^{n-1}}{\theta^n}$$

$$F_{Y_n}(y_n) = \frac{y_n^n}{\theta^n} \quad 0 < y < \theta$$

$$\text{Let } Z_n = Y_n/\theta$$

$$F_Z(z) = P(Z \le z) = P(Y_n \le Z\theta) = F_y(z\theta)$$

$$F_y(z\theta) = \frac{(z\theta)^n}{\theta^n} = z^n = F_Z(z)$$

$$f_Z(z) = nz^{n-1} \quad 0 < z < 1$$

$$Beta(n,1) = \frac{\Gamma(n+1)}{\Gamma(n)\Gamma(1)} z^{n-1} (1-z)^{1-1} \quad 0 < z < 1$$

$$= \frac{n!}{n-1!} z^{n-1} = nz^{n-1}$$

$$Z \sim Beta(n,1)$$

Problem 2

(a)

$$\bar{X} = \frac{1}{n} \sum_{i=1}^{n} X_i$$

$$\bar{Y} = \frac{1}{n} \sum_{i=1}^{n} Y_i$$

Since X_1, \ldots, X_n and Y_1, \ldots, Y_n are random samples from $N(\mu, \sigma^2)$:

$$\begin{split} \bar{X} \sim N(\mu, \sigma^2/n) \quad \bar{Y} \sim N(\mu, \sigma^2/n) \\ \bar{X} - \bar{Y} \sim N(\mu - \mu, \sigma^2/n + (-\sigma)^2/n) = N(0, 2\sigma^2/n) \end{split}$$

(b)

$$\begin{aligned} \text{WTS: } &\lim_{n \to \infty} P(|\bar{X} - \bar{Y}| > \sigma) = 0 \\ P(|\bar{X} - \bar{Y}| > \sigma) = 1 - P(|\bar{X} - \bar{Y}| \leq \sigma) \\ &= 1 - P(-\sigma \leq \bar{X} - \bar{Y} \leq \sigma) \\ &= 1 - P\left(\frac{-\sigma}{\sqrt{2}\sigma/\sqrt{n}} \leq \frac{\bar{X} - \bar{Y}}{\sqrt{2}\sigma/\sqrt{n}} \leq \frac{\sigma}{\sqrt{2}\sigma/\sqrt{n}}\right) \\ 1 - P\left(-\sqrt{n/2} \leq \frac{\bar{X} - \bar{Y}}{\sqrt{2}\sigma/\sqrt{n}} \leq \sqrt{n/2}\right) \\ &= 1 - (\Phi(\sqrt{n/2}) - \Phi(-\sqrt{n/2})) \\ &= 1 - \Phi(\sqrt{n/2}) + \Phi(-\sqrt{n/2})) \\ &\lim_{n \to \infty} 1 - \Phi(\sqrt{n/2}) + \Phi(-\sqrt{n/2}) \\ \lim_{n \to \infty} 1 - \Phi(\sqrt{n/2}) + \Phi(-\sqrt{n/2}) \\ \lim_{n \to \infty} \Phi(\sqrt{n/2}) = \Phi(\infty) = 1 \quad \lim_{n \to \infty} \Phi(-\sqrt{n/2}) = \Phi(-\infty) = 0 \\ &= 1 - 1 + 0 = 0 \\ \text{Thus } \lim_{n \to \infty} P(|\bar{X} - \bar{Y}| > \sigma) = 0 \end{aligned}$$

Problem 3

(a)

$$X_{1} \sim pois(\lambda_{1}) \quad X_{2} \sim pois(\lambda_{2})$$

$$X_{1} \perp X_{2}$$
WTS: $X_{1}|X_{1} + X_{2} \sim bin(n, p)$ where
$$E(X_{1}|X_{2} + X_{2} = n) = \frac{n\lambda_{1}}{\lambda_{1} + \lambda_{2}}$$

$$p = \frac{\lambda_{1}}{\lambda_{1} + \lambda_{2}}$$

$$(X_{1} + X_{2} = n) \sim pois(n|\lambda_{1} + \lambda_{2})$$

$$= \frac{e^{-(\lambda_{1} + \lambda_{2})}(\lambda_{1} + \lambda_{2})^{n}}{n!}$$

$$f_{X_{1}, X_{2}}(x_{1}, x_{2}) = \frac{e^{-(\lambda_{1} + \lambda_{2})}\lambda_{1}^{x_{1}}}{x_{1}!} \frac{\lambda_{2}^{x_{2}}}{x_{2}!}$$

$$P(X_{1} = x|X_{1} + X_{2} = n) = \frac{P(X_{1} = x, X_{1} + X_{2} = n)}{P(X_{1} + X_{2} = n)}$$

$$= \frac{P(X_1 = x)P(X_2 = n - x)}{P(X_1 + X_2 = n)}$$

$$= \frac{e^{-(\lambda_1 + \lambda_2)}\lambda_1^x \lambda_2^{n - x}}{x!(n - x)!} / \frac{(\lambda_1 + \lambda_2)^n e^{-(\lambda_1 + \lambda_2)}}{n!}$$

$$= \binom{n}{x} \frac{e^{-(\lambda_1 + \lambda_2)}\lambda_1^x \lambda_2^{n - x}}{(\lambda_1 + \lambda_2)^n e^{-(\lambda_1 + \lambda_2)}}$$

$$= \binom{n}{x} \frac{\lambda_1^x \lambda_2^{n - x}}{(\lambda_1 + \lambda_2)^n}$$

$$= \binom{n}{x} \frac{\lambda_1^x \lambda_2^{n - x}}{(\lambda_1 + \lambda_2)^n} \frac{(\lambda_1 + \lambda_2)^{n - x}}{(\lambda_1 + \lambda_2)^{n - x}}$$

$$= \binom{n}{x} \left(\frac{\lambda_2}{\lambda_1 + \lambda_2}\right)^{n - x} \left(\frac{\lambda_1}{\lambda_1 + \lambda_2}\right)^x$$
Which is $bin(n, \frac{\lambda_1}{\lambda_1 + \lambda_2})$
Thus $E(X_1 | X_2 + X_2 = n) = \frac{n\lambda_1}{\lambda_1 + \lambda_2}$

(b)

WTS: whether
$$T(X) = \sum_{i=1}^{2} X_i = X_1 + X_2$$
 is an SS for λ_1
$$P(X = x, T(X) = t) = \frac{P(X = x, X_1 + X_2 = t)}{P(X_1 + X_t)}$$
$$= \frac{P(X_1 = x, X_2 = x_2)}{P(X_1 + X_2 = t)}$$
$$= \frac{\lambda_1^{x_1} e^{-\lambda_1} \lambda_2^{x_2} e^{-\lambda_2}}{x_1! x_2!} / \frac{(\lambda_1 + \lambda_2)^t e^{-(\lambda_1 + \lambda_2)}}{t!}$$
$$= \frac{t!}{x_1! x_2!} \frac{\lambda_1^{x_1} \lambda_2^{x_2}}{(\lambda_1 + \lambda_2)^t}$$
This includes λ_1 Thus $(X_1 + X_2)$ not $\pm \lambda_1$ $X_1 + X_2$ not SS of λ_1

Problem 4

(a)

$$f_X(x) = \frac{x^{\beta - 1}e^{-x}}{\Gamma(\beta)} \quad x > 0, \quad \beta > 1$$

$$f_T(t|X = x) = \theta x e^{-\theta x t} \quad t > 0, \quad \theta > 0$$

$$f_T(t) = \int_0^\infty f_{T,X}(t,x) \, dx$$

$$= \int_0^\infty f_T(t|X = x) f_X(x) \, dx$$

$$= \int_0^\infty \theta x e^{-\theta x t} \frac{x^{\beta - 1}e^{-x}}{\Gamma(\beta)} \, dx$$

$$= \frac{\theta}{\Gamma(\beta)} \int_0^\infty x^\beta e^{-(\theta t + 1)x} \, dx$$

Want to get integral in the form of $gamma(\beta + 1, 1/(\theta t + 1))$

$$= \frac{\Gamma(\beta+1)\theta}{\Gamma(\beta)(\theta t+1)^{\beta+1}} \int_0^\infty \frac{(\theta t+1)^{\beta+1}}{\Gamma(\beta+1)} x^\beta e^{-(\theta t+1)x} dx$$
$$= \frac{\Gamma(\beta+1)\theta}{\Gamma(\beta)(\theta t+1)^{\beta+1}}$$
$$f_T(t) = \frac{\beta \theta}{(\theta t+1)^{\beta+1}}$$

(b)

WTS:
$$E(T) = \frac{1}{\theta(\beta - 1)}$$

 $E(T) = E(E(T|X = x))$
 $E(T|X = x) = 1/(\theta X)$ giving us:
 $= E(1/(\theta X)) = \frac{1}{\theta}E(X^{-1})$
 $E(X^{-1}) = \int_0^\infty x^{-1} f_X(x) dx$
 $= \int_0^\infty \frac{x^{\beta - 2} e^{-x}}{\Gamma(\beta)} dx$
 $= \frac{\Gamma(\beta - 1)}{\Gamma(\beta)} \int_0^\infty \frac{x^{\beta - 2} e^{-x}}{\Gamma(\beta - 1)} dx$
 $= \frac{\Gamma(\beta - 1)}{\Gamma(\beta)}$
 $= \frac{(\beta - 2)!}{(\beta - 1)!}$

$$E(T) = \frac{1}{\theta}E(X^{-1}) = \frac{1}{\theta(\beta - 1)}$$

(c)

Given
$$\beta = 2$$
 $E(T) = \frac{1}{\theta(2-1)} = \frac{1}{\theta}$ and $\sqrt{n}(\hat{\theta} - \theta) \stackrel{d}{\to} N(0, 2\theta^2)$ $T_n = \hat{\theta}$ Using delta method: $\sqrt{n}\{g(T_n) - g(\theta)\} \stackrel{d}{\to} N(0, \{g^{'}(\theta)\}^2 \sigma^2)$
$$g(\theta) = \frac{1}{\theta} \quad g^{'}(\theta) = -\theta^{-2}$$

$$\sqrt{n}(1/\hat{\theta} - 1/\theta) \stackrel{d}{\to} N(0, \theta^{-4}2\theta^2) = N(0, 2\theta^{-2})$$

$$\sigma^2 = 2\theta^{-2}$$