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2. Solution:

Proof. Since X_1, \ldots, X_n are independent exponential random variables with parameter λ , then their joint pdf is

$$f_X(x_1, \dots, x_n) = f(x_1) \dots f(x_N)$$

$$= \lambda e^{-\lambda x_1} \dots \lambda e^{-\lambda x_n}$$

$$= \begin{cases} \lambda^n e^{-\lambda \sum_{i=1}^n x_i}, & \text{if } (x_1, \dots, x_n) \in \mathbb{R}^n_{\geq 0} \\ 0, & \text{otherwise} \end{cases}$$

Also, the joint pdf of $X_{(1)}, \ldots, X_{(n)}$ is

$$f_{1,\dots,n}(x_1,\dots,x_n) = n! f(x_1) \dots f(x_n)$$

$$= \begin{cases} n! \lambda^n e^{-\lambda \sum_{i=1}^n x_i}, & \text{if } 0 < x_1 < \dots < x_n. \\ 0, & \text{otherwise.} \end{cases}$$

Given the above facts, let's find the joint pdf of Y_1, \ldots, Y_n . For $i \in \{1, \ldots, n\}$, define the following set of functions:

$$g_i(y_1, \dots, y_n) = \sum_{k=1}^{i} \frac{y_i}{n+1-k}$$

We then have

$$|J_g(y_1, \dots, y_n)| = \begin{vmatrix} \frac{\partial g_1}{\partial y_1} & \dots & \frac{\partial g_1}{\partial y_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial g_n}{\partial y_1} & \dots & \frac{\partial g_n}{\partial y_n} \end{vmatrix}$$

$$= \begin{vmatrix} \frac{1}{n} & & & & \\ \frac{1}{n} & \frac{1}{n-1} & & & \\ \frac{1}{n} & \frac{1}{n-1} & \frac{1}{n-2} & & & \\ \vdots & & & \frac{1}{2} & \\ \frac{1}{n} & & & & \frac{1}{2} & \frac{1}{1} \end{vmatrix}$$

$$= \prod_{i=1}^n \frac{1}{i} \qquad \qquad \therefore J_g \text{ is a lower triangular matrix.}$$

$$= \frac{1}{n!}$$

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For the support of $Y = (Y_1, \ldots, Y_n)$,

$$0 < \frac{y_1}{n} < \frac{y_1}{n} + \frac{y_2}{n-1} < \dots < \frac{y_1}{n} + \dots + y_n$$

$$\Rightarrow (y_1, \dots, y_n) \in \mathbb{R}^n_{>0}$$

It follows that the joint pdf of Y_1, \ldots, Y_n is

$$f_{Y}(y_{1},...,y_{n}) = f_{1,...,n}(g_{1}(y_{1},...,y_{n}),...,g_{n}(y_{1},...,y_{n}))|J_{g}(y_{1},...,y_{n})|$$

$$= n!\lambda^{n} \left(e^{-\lambda \sum_{i=1}^{n} \sum_{k=1}^{i} \frac{y_{i}}{n+1-k}}\right) \frac{1}{n!}$$

$$= \begin{cases} \lambda^{n}e^{-\lambda \sum_{i=1}^{n} y_{i}}, & \text{if } (y_{1},...,y_{n}) \in \mathbb{R}^{n}_{\geq 0} \\ 0, & \text{otherwise} \end{cases}$$

$$= f_{X}(y_{1},...,y_{n})$$

Therefore, Y_1, \ldots, Y_n has the same joint distribution as X_1, \ldots, X_n .

For $i \in \{1, ..., n\}$ and $\{j_1, ..., j_{n-1}\} = \{1, ..., n\} \setminus \{i\}$, the marginal pdf for Y_i is

$$f_{Y_i}(y_i) = \int \cdots \int f_Y(y_1, \dots, y_n) \, \mathrm{d}y_{j_1} \dots \, \mathrm{d}y_{j_{n-1}}$$

$$= \int_0^\infty \cdots \int_0^\infty \lambda^n e^{-\lambda \sum_{k=1}^n y_k} \, \mathrm{d}y_{j_1} \dots \, \mathrm{d}y_{j_{n-1}}$$

$$= \begin{cases} \lambda e^{-\lambda y_i}, & \text{if } y_i \in \mathbb{R}_{\geq 0} \\ 0, & \text{otherwise} \end{cases}$$

It can be seen that $f_Y(y_1, ..., y_n) = f_{Y_1}(y_1) ... f_{Y_n}(y_n)$.

Hence, Y_1, \ldots, Y_n are also independent.