

(1) The joint density of X and Y is given by

$$f(x, y) = \begin{cases} \frac{1}{y} e^{-x/y - y}, & 0 < x, y < \infty; \\ 0, & \text{otherwise.} \end{cases}$$

We have seen how to calculate $\mathbb{E}(X)$, $\mathbb{E}(Y)$, $\mathbb{E}(XY)$, and also $Cov(X, Y)$ from last tutorial. Now find $\mathbb{P}\{X > 1 | Y = y\}$ for $y > 0$.

Need to find conditional PDF of $f(x|y)$

$$f(x|y) = \frac{f(x, y)}{f_Y(y)} = \frac{\cancel{\frac{1}{y}} e^{-\frac{x}{y} - y}}{\int_0^\infty \cancel{\frac{1}{y}} e^{-\frac{x}{y} - y} dx} = \frac{e^{-\frac{x}{y}}}{\left[-y e^{-\frac{x}{y}}\right]_{x=0}^{x=\infty}} = \frac{e^{-\frac{x}{y}}}{0 + y} = \frac{1}{y} e^{-\frac{x}{y}}$$

$$\mathbb{P}(X > 1 | Y = y) = \int_1^\infty f(x|y) dx = \frac{1}{y} \int_1^\infty e^{-\frac{x}{y}} dx = \frac{1}{y} \left[y e^{-\frac{x}{y}} \right]_1^\infty = e^{-\frac{1}{y}}$$

(2) The joint density of X and Y is given by

$$f(x, y) = \frac{1}{y} e^{-y}, \quad 0 < x < y < \infty.$$

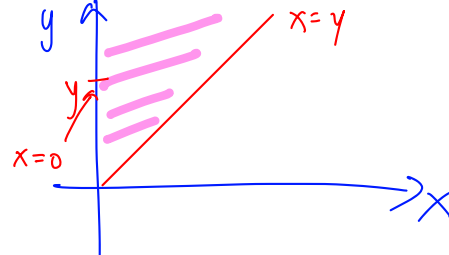
Find $\mathbb{E}[X^3 | Y = y]$

$$\mathbb{E}[X^3 | Y = y] = \int_0^y x^3 f(x|y) dx$$

For a given $Y=y$, x changes from $x=0$ to $x=y$

$$f(x|y) = \frac{f(x, y)}{f_Y(y)} = \frac{\cancel{\frac{1}{y}} e^{-y}}{\int_0^y \cancel{\frac{1}{y}} e^{-y} dx} = \frac{1}{y}$$

$$\mathbb{E}[X^3 | Y=y] = \int_0^y x^3 \cdot \frac{1}{y} dx = \frac{1}{4} y^3$$



(3) The joint density of X and Y is given by

$$f(x, y) = \begin{cases} e^{-(x+y)}, & 0 < x, y < \infty \\ 0, & \text{otherwise.} \end{cases}$$

Find the density function of the random variable $Z = X/Y$.

CDF of Z : $F_Z(z) = P(Z \leq z) = P\left(\frac{X}{Y} \leq z\right) = P(X \leq Yz)$

$$= \int_0^\infty \int_0^{yz} e^{-x-y} dx dy \quad (z > 0)$$

\swarrow
 $f(x, y)$

Fundamental Thm of Calc.

$$\frac{d}{dx} \int_{\text{const.}}^x f(t) dt = f(x)$$

PDF of Z

$$f_Z(z) = \frac{d}{dz} F_Z(z)$$

$$= \int_0^\infty \frac{d}{dz} \left(\int_0^{yz} e^{-x-y} dx \right) dy$$

Let $w = yz$

$$= \int_0^\infty \frac{d}{dw} \left(\int_0^w e^{-x-y} dx \right) \cdot \frac{\partial w}{\partial z} dy$$

$$= \int_0^\infty e^{-w-y} \cdot y dy$$

Int by parts

$$= \int_0^\infty y e^{-y(1+z)} dy$$

$$= \left[-\frac{y}{1+z} e^{-y(1+z)} \right]_0^\infty + \frac{1}{1+z} \int_0^\infty e^{-y(1+z)} dy$$

$= 0$

$$= \frac{1}{z+1} \left[-\frac{1}{z+1} e^{-y(1+z)} \right]_0^\infty = \frac{1}{(z+1)^2}$$

$$f_Z(z) = \begin{cases} \frac{1}{(z+1)^2} & z > 0 \\ 0 & \text{else.} \end{cases}$$

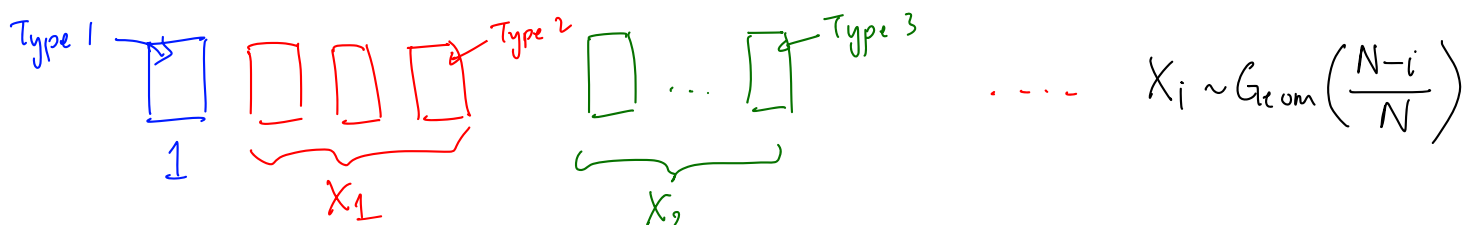
$$Z = \frac{X}{Y}$$

- (4) In the coupon collector problem, show that the variance of the number of coupons needed to amass a full set is equal to

$$\sum_{i=1}^{N-1} \frac{iN}{(N-i)^2}.$$

When N is large, this can be shown to be approximately equal (in the sense that their ratio approaches 1 as $N \rightarrow \infty$) to $N^2\pi^2/6$.

Let X_i be the number of additional steps to see the $(i+1)^{\text{th}}$ type after having collected distinct i types of coupons.



Let X be the total number of coupons collected,

$$X = 1 + X_1 + X_2 + \dots + X_{N-1}$$

$$\text{Var}(X) = \text{Var}(1 + X_1 + \dots + X_{N-1})$$

$$= \text{Var}\left(\sum_{i=1}^{N-1} X_i\right)$$

$$= \sum_{i=1}^{N-1} \text{Var}(X_i) \quad (X_i\text{'s are indep.})$$

$$= \sum_{i=1}^{N-1} \frac{1 - \left(\frac{N-i}{N}\right)}{\left(\frac{N-i}{N}\right)^2} \quad (\text{Var of } \text{Geo}\left(\frac{N-i}{N}\right))$$

$$= \sum_{i=1}^{N-1} \frac{iN}{(N-i)^2}$$

$$= \sum_{i=1}^{N-1} \frac{N^2}{(N-i)^2} - \sum_{i=1}^{N-1} \frac{N(N-i)}{(N-i)^2}$$

$$\text{Write } iN = N^2 - N(N-i)$$

$$= N^2 \sum_{i=1}^{N-1} \frac{1}{(N-i)^2} - N \sum_{i=1}^{N-1} \frac{1}{N-i} = N^2 \sum_{i=1}^{N-1} \frac{1}{i^2} - N \sum_{i=1}^{N-1} \frac{1}{i}$$

$$N \rightarrow \infty \quad \sum_{i=1}^N \frac{1}{i^2} \rightarrow \frac{\pi^2}{6} \quad \sum_{i=1}^N \frac{1}{i} \rightarrow \ln N + \gamma \quad \leftarrow \text{Euler's constant } 0.577\dots$$

(5) Let X and Y be independent uniform random variables on $(0,1)$. Find the density function of $Z = X + Y$.

$$\begin{aligned} \text{CDF of } Z : F_Z(z) &= P(X + Y \leq z) \\ &= P(Y \leq z - X) \end{aligned}$$

$$= \int_0^1 \int_0^{z-x} f(x, y) dy dx$$

$$f_X(x) = \begin{cases} 1 & 0 < x < 1 \\ 0 & \text{otherwise} \end{cases}$$

$$f_Y(y) = \begin{cases} 1 & 0 < y < 1 \\ 0 & \text{otherwise} \end{cases}$$

$$\text{PDF of } Z : f_Z(z) = \frac{d}{dz} F_Z(z) = \int_0^1 \left[\frac{d}{dz} \int_0^{z-x} f(x, y) dy \right] dx$$

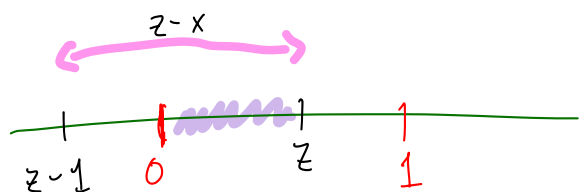
$$= \int_0^1 f(x, z-x) dx$$

$$\begin{aligned} &\text{Convolution product} \left[\int_0^1 \underbrace{f_X(x)}_{=1} f_Y(z-x) dx \right] = \int_0^1 f_Y(z-x) dx \end{aligned}$$

$0 < z < 2$

For $f_Y(z-x) = 1$, we need $z-x \in (0,1)$ while $0 < x < 1$

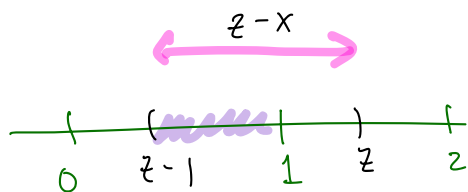
z is fixed but unknown



when $0 < z < 1$

$$f_Z(z) = \int_0^1 f_Y(z-x) dx = \int_0^z 1 dx = z$$

x changes from 0 to 1.



when $1 < z < 2$

$$f_Z(z) = \int_0^1 f_Y(z-x) dx = \int_{z-1}^1 1 dx = 2 - z$$

(6) Suppose that X_1, \dots, X_n are i.i.d. $\text{Exp}(\lambda)$ random variables. Prove that $X_1 + \dots + X_n \sim \text{Gamma}(n, \lambda)$.

$\text{Exp}(\lambda)$ is same as $\text{Gamma}(1, \lambda)$

It is sufficient to show if $X_1 \sim \text{Gamma}(\alpha, \lambda)$
 $X_2 \sim \text{Gamma}(\beta, \lambda)$, then $X_1 + X_2 \sim \text{Gamma}(\alpha + \beta, \lambda)$

$$f_{X_1 + X_2}(z) = \int_{-\infty}^{\infty} f_{X_1}(x) f_{X_2}(z-x) dx$$

$$= \int_0^z \frac{\lambda^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{-\lambda x} \cdot \frac{\lambda^\beta}{\Gamma(\beta)} (z-x)^{\beta-1} e^{-\lambda(z-x)} dx$$

$$= \frac{\lambda^\alpha \lambda^\beta}{\Gamma(\alpha) \Gamma(\beta)} e^{-\lambda z} \int_0^z x^{\alpha-1} (z-x)^{\beta-1} dx$$

$$= \frac{\lambda^{\alpha+\beta}}{\Gamma(\alpha) \Gamma(\beta)} e^{-\lambda z} \int_0^1 (uz)^{\alpha-1} (z-uz)^{\beta-1} \cdot z du$$

let $u = \frac{x}{z}$

$$= \frac{\lambda^{\alpha+\beta}}{\Gamma(\alpha) \Gamma(\beta)} e^{-\lambda z} z^{\alpha+\beta-1} \int_0^1 u^{\alpha-1} (1-u)^{\beta-1} du$$

$$= \frac{\lambda^{\alpha+\beta}}{\Gamma(\alpha+\beta)} e^{-\lambda z} z^{\alpha+\beta-1}$$

Beta function, $B(\alpha, \beta) = \frac{\Gamma(\alpha) \Gamma(\beta)}{\Gamma(\alpha+\beta)}$

$$\therefore X_1 + X_2 \sim \text{Gamma}(\alpha + \beta, \lambda)$$

Proof of $\frac{\Gamma(\alpha) \Gamma(\beta)}{\Gamma(\alpha+\beta)} = \int_0^1 x^{\alpha-1} (1-x)^{\beta-1} dx$

$$\begin{aligned}\Gamma(\alpha) \Gamma(\beta) &= \int_0^\infty e^{-x} x^{\alpha-1} dx \cdot \int_0^\infty e^{-y} y^{\beta-1} dy \\ &= \int_0^\infty \int_0^\infty e^{-(x+y)} x^{\alpha-1} y^{\beta-1} dx dy\end{aligned}$$

Let $x=st$, $y=s(1-t)$ $dx dy = \begin{vmatrix} \frac{\partial x}{\partial t} & \frac{\partial x}{\partial s} \\ \frac{\partial y}{\partial t} & \frac{\partial y}{\partial s} \end{vmatrix} dt ds$

$$= \begin{vmatrix} s & t \\ -s & 1-t \end{vmatrix} dt ds$$

$$= s \cdot dt ds$$

$$\begin{aligned}\Gamma(\alpha) \Gamma(\beta) &= \int_0^\infty \int_0^1 e^{-s} (st)^{\alpha-1} (s(1-t))^{\beta-1} \cdot s dt ds \\ &= \int_0^\infty e^{-s} \cdot s^{\alpha+\beta-1} ds \int_0^1 t^{\alpha-1} (1-t)^{\beta-1} dt \\ &= \Gamma(\alpha+\beta) \cdot \int_0^1 t^{\alpha-1} (1-t)^{\beta-1} dt.\end{aligned}$$