

Sums of independent random variables

Q. X, Y independent random variables. What is the distribution of $X+Y$?

Let's find the cdf when X and Y are (jointly) continuous:

$$F_{X+Y}(a) = \mathbb{P}(X+Y \leq a) = \iint_{x+y \leq a} f(x,y) dx dy$$

$$= \iint_{x+y \leq a} f_X(x) f_Y(y) dx dy$$

independence

$$= \int_{-\infty}^{\infty} \int_{-\infty}^{a-y} f_X(x) dx f_Y(y) dy$$

$$= \int_{-\infty}^{\infty} F_X(a-y) f_Y(y) dy$$

~~Step~~ Differentiate (in a):

$$f_{X+Y}(a) = \int_{-\infty}^{\infty} f_X(a-y) f_Y(y) dy$$

"convolution of f_X and f_Y "

Ex. $X, Y \sim U(0,1)$ independent.

$$f_{X+Y}(a) = \int_{-\infty}^{\infty} f_X(a-y) f_Y(y) dy$$

$$= \int_0^1 f_X(a-y) dy$$

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

$$= 0 \text{ unless}$$

$$; f_X(a-y) = \begin{cases} 0 & \text{unless} \\ 1 & 0 < a-y < 1 \end{cases}$$

\downarrow

$$a-1 < y < a$$

~~NA~~

If $a < 0$, $a-1 \leq y \leq a$ implies $y < 0$. So

$$\int_0^1 f_X(a-y) dy = 0$$

If $0 \leq a \leq 1$, $\left. \begin{array}{l} a-1 \leq y \leq a \\ 0 \leq y \leq 1 \end{array} \right\} \iff 0 \leq y \leq a$. In this case,

$$\int_0^1 f_X(a-y) dy = \int_0^a f dy = a$$

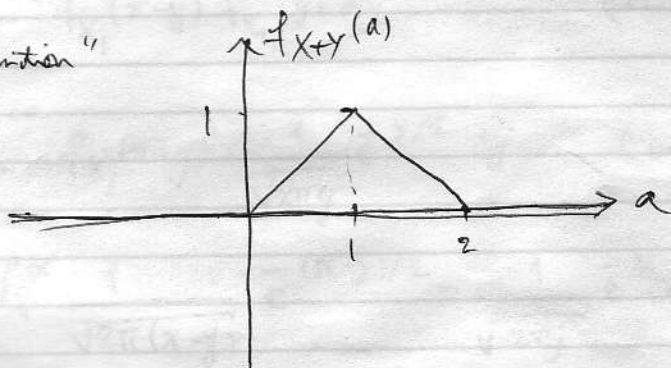
If $1 < a < 2$, $\left. \begin{array}{l} a-1 \leq y \leq a \\ 0 \leq y \leq 1 \end{array} \right\} \iff a-1 \leq y \leq 1$. Here,

$$\int_0^1 f_X(a-y) dy = \int_{a-1}^1 dy = 1 - (a-1) = 2-a$$

If $a > 2$, $\int_0^1 f_X(a-y) dy = 0$.

$$\therefore f_{X+Y}(a) = \begin{cases} a & , \quad 0 \leq a \leq 1 \\ 2-a & , \quad 1 \leq a \leq 2 \\ 0 & , \quad \text{otherwise} \end{cases}$$

"triangular density function"



Example, X_1, X_2 independent $N(0,1)$'s.

(a) Find $f_{X_1^2}$.

(b) Find $f_{X_1^2+X_2^2}$.

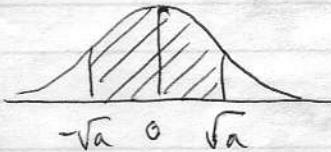
Solⁿ to (a)

~~Find~~ $\mathbb{P}(X_1^2 \leq a) = \mathbb{P}(-\sqrt{a} \leq X_1 \leq \sqrt{a})$

$a > 0$

$$= \Phi(\sqrt{a}) - \Phi(-\sqrt{a})$$

$$= 2\Phi(\sqrt{a}) - 1$$



$$\therefore f_{X_1^2}(a) = 2 \cdot \frac{1}{2\sqrt{a}} \Phi'(\sqrt{a})$$

$$= \frac{1}{\sqrt{2\pi a}} e^{-a/2} \quad a > 0$$

If $a < 0$, $f_{X_1^2}(a) = 0$, since $\mathbb{P}(X_1^2 \leq a) = 0$ then.

Solⁿ to (b)

$$f_{X_1^2+X_2^2}(x) = \int_{-\infty}^{\infty} f_{X_1}(x-y) f_{X_2}(y) dy \quad (\text{convolution})$$

$$= \int_0^{\infty} f_{X_1}(x-y) \frac{1}{\sqrt{2\pi y}} e^{-y/2} dy \quad (\text{part (a)})$$

$$= \int_0^x \frac{1}{\sqrt{2\pi(x-y)}} e^{-(x-y)/2} \cdot \frac{1}{\sqrt{2\pi y}} e^{-y/2} dy \quad (\text{part (a)})$$

$$= \frac{e^{-x/2}}{2\pi} \int_0^x \frac{1}{\sqrt{y(x-y)}} dy$$

$$\sqrt{y(x-y)} = x \sqrt{\frac{y}{x} \left(1 - \frac{y}{x}\right)}, \text{ since } x > 0. \text{ Therefore,}$$

$$f_{X_1^2 + X_2^2}(x) = \frac{e^{-x/2}}{2\pi x} \int_0^x \frac{dy}{\sqrt{\frac{y}{x} \left(1 - \frac{y}{x}\right)}}$$

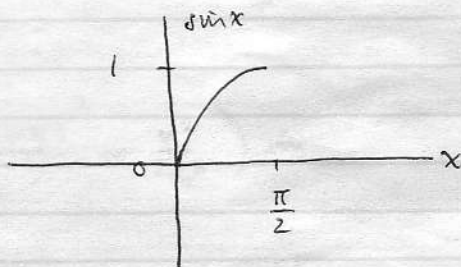
$$= \frac{e^{-x/2}}{2\pi} \int_0^1 \frac{dz}{\sqrt{z(1-z)}} \quad (z = y/x)$$

$$= \frac{e^{-x/2}}{\pi} \int_0^1 \frac{dw}{\sqrt{1-w^2}} \quad (w = \sqrt{z})$$

$$= \frac{e^{-x/2}}{\pi} (\sin^{-1} 1 - \sin^{-1} 0)$$

$$= \frac{e^{-x/2}}{\pi} \left(\frac{\pi}{2} - 0\right)$$

$$= \frac{1}{2} e^{-x/2}$$



⇒

$$f_{X_1^2 + X_2^2}(x) = \begin{cases} \frac{1}{2} e^{-x/2} & , x > 0 \\ 0 & , x \leq 0 \end{cases}$$

ie. $\boxed{X_1^2 + X_2^2 \sim \text{Exp}\left(\frac{1}{2}\right)}$

Sums of independent normals.

4// X_1, X_2 indep't. $N(\mu_1, \sigma_1^2)$, $N(\mu_2, \sigma_2^2)$, resp.
 Find the pdf of $X_1 + X_2$.

$$\begin{aligned} f_{X_1+X_2}(a) &= \int_{-\infty}^{\infty} f_{X_1}(a-u) f_{X_2}(u) du \\ &= \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}\sigma_1} \exp\left(-\frac{(a-u-\mu_1)^2}{2\sigma_1^2}\right) \frac{1}{\sqrt{2\pi}\sigma_2} \exp\left(-\frac{(u-\mu_2)^2}{2\sigma_2^2}\right) du \\ &= \int_{-\infty}^{\infty} \frac{1}{2\pi\sigma_1\sigma_2} \exp\left(-\frac{(a-u-\mu_1)^2}{2\sigma_1^2} - \frac{(u-\mu_2)^2}{2\sigma_2^2}\right) du \end{aligned}$$

~~$\frac{1}{2\pi\sigma_1\sigma_2} \exp\left(-\frac{(a-u-\mu_1)^2}{2\sigma_1^2} - \frac{(u-\mu_2)^2}{2\sigma_2^2}\right) du$~~ $v = u - \mu_2$

$$= \int_{-\infty}^{\infty} \frac{1}{2\pi\sigma_1\sigma_2} \exp\left(-\frac{(a-(\mu_1+\mu_2)-v)^2}{2\sigma_1^2} - \frac{v^2}{2\sigma_2^2}\right) dv$$

let $\mu = \mu_1 + \mu_2$, for simplicity.

$$f_{X_1+X_2}(a) = \int_{-\infty}^{\infty} \frac{1}{2\pi\sigma_1\sigma_2} \exp\left(-\frac{(a-\mu-v)^2}{2\sigma_1^2} - \frac{v^2}{2\sigma_2^2}\right) dv$$

$$\frac{(a-\mu-v)^2}{2\sigma_1^2} + \frac{v^2}{2\sigma_2^2} = \frac{(a-\mu)^2 + v^2 - 2(a-\mu)v}{2\sigma_1^2} + \frac{v^2}{2\sigma_2^2}$$

$$= \frac{v^2}{2} \left(\frac{1}{\sigma_1^2} + \frac{1}{\sigma_2^2} \right) + \frac{(a-\mu)^2}{2\sigma_1^2} - \frac{(a-\mu)v}{\sigma_1^2}$$

Let $\sigma^2 = \sigma_1^2 + \sigma_2^2$, for simplicity.

$$\frac{(a-\mu-v)^2}{2\sigma_1^2} + \frac{v^2}{2\sigma_2^2} = \frac{v^2}{2} \cdot \frac{\sigma^2}{\sigma_1^2\sigma_2^2} - 2\left(\frac{a-\mu}{2\sigma_1^2}\right)v + \frac{(a-\mu)^2}{2\sigma_1^2}$$

~~$\frac{1}{2\pi\sigma_1\sigma_2} \exp\left(-\frac{(a-\mu-v)^2}{2\sigma_1^2} - \frac{v^2}{2\sigma_2^2}\right) dv$~~

$$= \frac{1}{2\sigma_1^2} \left[v^2 \cdot \frac{\sigma^2}{\sigma_2^2} - 2(a-\mu)v \right] + \frac{(a-\mu)^2}{2\sigma_1^2}$$

$$\begin{aligned}
 &= \frac{1}{2\sigma_1^2} \left[v^2 \cdot \frac{\sigma_2^2}{\sigma_2^2} - 2(a-\mu) \frac{\sigma_2}{\sigma} \left(v \frac{\sigma}{\sigma_2} \right) \right] + \frac{(a-\mu)^2}{2\sigma_1^2} \\
 &= \frac{1}{2\sigma_1^2} \left[\left(\frac{v\sigma}{\sigma_2} \right)^2 - 2(a-\mu) \frac{\sigma_2}{\sigma} \left(v \frac{\sigma}{\sigma_2} \right) + (a-\mu)^2 \frac{\sigma_2^2}{\sigma^2} \right] - \frac{(a-\mu)^2 \sigma_2^2}{2\sigma^2 \sigma_1^2} + \frac{(a-\mu)^2}{2\sigma_1^2} \\
 &= \frac{1}{2\sigma_1^2} \left(\frac{v\sigma}{\sigma_2} - (a-\mu) \frac{\sigma_2}{\sigma} \right)^2 + \frac{(a-\mu)^2}{2\sigma_1^2} \left[1 - \frac{\sigma_2^2}{\sigma^2} \right] \\
 &= \frac{1}{2\sigma_1^2} \left[\frac{v\sigma}{\sigma_2} - \frac{(a-\mu)\sigma_2}{\sigma} \right]^2 + \frac{(a-\mu)^2}{2\sigma^2} \quad \left[\sigma^2 \triangleq \sigma_1^2 + \sigma_2^2 \right]
 \end{aligned}$$

$$\begin{aligned}
 \therefore f_{X_1+X_2}(a) &= \int_{-\infty}^{\infty} \frac{1}{2\pi\sigma_1\sigma_2} \exp \left[-\frac{1}{2\sigma_1^2} \left(\frac{v\sigma}{\sigma_2} - \frac{(a-\mu)\sigma_2}{\sigma} \right)^2 \right] dv \cdot e^{-\frac{(a-\mu)^2}{2\sigma^2}} \\
 &= \int_{-\infty}^{\infty} \frac{1}{2\pi\sigma_1\sigma_2} e^{-w^2/2\sigma_1^2} dw \cdot \frac{\sigma_2}{\sigma} \cdot \exp \left[-\frac{(a-\mu)^2}{2\sigma^2} \right] \quad \left[w = \frac{v\sigma}{\sigma_2} - \dots \right] \\
 &= \frac{1}{\sqrt{2\pi}} \exp \left[-\frac{(a-\mu)^2}{2\sigma^2} \right].
 \end{aligned}$$

$$\Rightarrow X_i \sim N(\mu_i, \sigma_i^2) \text{ indep.} \Rightarrow X_1 + X_2 \sim N(\mu_1 + \mu_2, \sigma_1^2 + \sigma_2^2).$$

$$\Rightarrow \underbrace{X_1 + X_2 + X_3}_{N(\mu_1 + \mu_2, \sigma_1^2 + \sigma_2^2 + \sigma_3^2)} \sim N(\mu_1 + \mu_2 + \mu_3, \sigma_1^2 + \sigma_2^2 + \sigma_3^2)$$

Theorem. $X_i \sim N(\mu_i, \sigma_i^2)$, $i=1, \dots, m$, independent.

$$\text{Then } \sum_{j=1}^m X_j \sim N\left(\sum_{i=1}^m \mu_i, \sum_{i=1}^m \sigma_i^2\right).$$

Summing Independent r.v.'s in the discrete case.

X_1, X_2 independent.

$$\mathbb{P}(X=x, Y=y) = p(x,y) = \mathbb{P}(X=x) \mathbb{P}(Y=y) = P_X(x) P_Y(y).$$

$$p(k) = \mathbb{P}_{X_1+X_2}(X_1+X_2=k) = \mathbb{P}(X_1=0, X_2=k) + \mathbb{P}(X_1=1, X_2=k-1) \dots \left. \vphantom{p(k)} \right\} \text{ pos. integer case}$$
$$= P_{X_1}(0) P_{X_2}(k) + P_{X_1}(1) P_{X_2}(k-1) + \dots$$

In general,

$$\mathbb{P}(X_1+X_2=k) = \sum_x P_{X_1}(x) P_{X_2}(k-x)$$

Example. X_1, X_2 indep't $X_i \sim \text{Pois}(\lambda_i) \quad i=1,2.$

$\mathbb{P}(X_1+X_2=k) = 0$, unless $k=0,1,2,\dots$ in which case,

$$\begin{aligned} \mathbb{P}(X_1+X_2=k) &= \sum_x P_{X_1}(x) P_{X_2}(k-x) \\ &= \sum_{x=0}^{\infty} \frac{e^{-\lambda_1} \lambda_1^x}{x!} P_{X_2}(k-x) \\ &= \sum_{x=0}^k \frac{e^{-\lambda_1} \lambda_1^x}{x!} \frac{e^{-\lambda_2} \lambda_2^{k-x}}{(k-x)!} \\ &= \frac{e^{-(\lambda_1+\lambda_2)}}{k!} \sum_{x=0}^k \binom{k}{x} \lambda_1^x \lambda_2^{k-x} \\ &= \frac{e^{-(\lambda_1+\lambda_2)}}{k!} (\lambda_1+\lambda_2)^k \quad (\text{Binomial Th}^m) \end{aligned}$$

$$\Rightarrow X_1+X_2 \sim \text{Pois}(\lambda_1+\lambda_2).$$

$$\Rightarrow \text{If } X_i \sim \text{Pois}(\lambda_i) \text{ indep't. , } \sum_{j=1}^n X_j \sim \text{Pois}(\sum_{j=1}^n \lambda_j).$$

Ex. X_1, X_2 indept. $X_i \sim \text{Bin}(\overset{n_i, p}{\cancel{n_i, p}})$ $i=1, 2$ (same p)

$X_1 + X_2 = \#$ of success in $n_1 + n_2$ ^{indept} trials $P(S) = p$

$\Rightarrow X_1 + X_2 \sim \text{Bin}(n_1 + n_2, p)$. More generally, $\sum_{j=1}^N X_j \sim \text{Bin}(\sum_{j=1}^N n_j, p)$.

Another calculation: $\mathbb{P}(X_1 + X_2 = k) = 0$, unless $k = 0, \dots, n_1 + n_2$, in which case,

$$\mathbb{P}(X_1 + X_2 = k) = \sum_x \mathbb{P}(X_1 = x) \mathbb{P}(X_2 = k - x)$$

$$= \sum_{x=0}^k \binom{n_1}{x} p^x (1-p)^{n_1-x} \binom{n_2}{k-x} p^{k-x} (1-p)^{n_2-(k-x)}$$

~~$$= \sum_{x=0}^k \frac{n_1!}{x!(n_1-x)!} p^x (1-p)^{n_1-x} \frac{n_2!}{(k-x)!(n_2-(k-x))!} p^{k-x} (1-p)^{n_2-(k-x)}$$~~

$$= p^k (1-p)^{n_1+n_2-k} \underbrace{\sum_{x=0}^k \binom{n_1}{x} \binom{n_2}{k-x}}_{\text{# of ways to pick from } n_1+n_2 \text{ things } k \text{ things - unordered} = \binom{n_1+n_2}{k}}$$

$$\mathbb{P}(X_1 + X_2 = k) = \begin{cases} \binom{n_1+n_2}{k} p^k (1-p)^{n_1+n_2-k} & , k=0, \dots, n_1+n_2 \\ 0 & , \text{o/w} \end{cases}$$