

Last time

Sec 5.3 independent normal variables

a) $T \sim \text{Exp}(\frac{1}{2}) \Rightarrow R = \sqrt{T}$ has density $f(r) = r e^{-\frac{1}{2}r^2}, r > 0$
 ↪ called Rayleigh RV

b) Using Rayleigh distribution we showed if $Z_1, Z_2 \stackrel{iid}{\sim} N(0,1)$

- ① $Z_1^2 + Z_2^2 = T \sim \text{Exp}(\frac{1}{2}),$
- ② $\Phi(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}}, E(z) = 0, \text{SD}(z) = 1$

c) Definition
 If $Z \sim N(0,1)$ then $X = \mu + \sigma Z \sim N(\mu, \sigma^2)$
 with density $f(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}$

d)

Then let $X_1 \sim N(\mu_1, \sigma_1^2)$
 $X_2 \sim N(\mu_2, \sigma_2^2)$ } indep.

then $aX_1 + bX_2 \sim N(a\mu_1 + b\mu_2, a^2\sigma_1^2 + b^2\sigma_2^2)$

Today Sec 5.3

① A linear combination of independent normals is normal

② Chi square distribution

Sec 5.4

③ Convolution formula for the density of $X+Y$

tinyurl

<http://tinyurl.com/april12-pt1>

<http://tinyurl.com/april12-pt2>

Stat 134

Friday November 2 2018

1. Let $X \sim N(0, 1)$ and $Y \sim N(0, 1)$ be independent. $P(X > 2Y)$ equals

- a** $1 - \Phi(0)$
- b** $1 - \Phi\left(\frac{1}{\sqrt{3}}\right)$
- c** $1 - \Phi(\sqrt{3})$
- d** none of the above

Solu
— $Y \sim N(0, 1) \Rightarrow ZY \sim N(0, 1)$

and $X - ZY \sim N(0, 5)$

$$P(X > ZY) = P(X - ZY > 0)$$

normalize $\xrightarrow{\text{normalize}} = P\left(\frac{(X - ZY) - 0}{\sqrt{5}} > \frac{0 - 0}{\sqrt{5}}\right)$

$$= P(Z > 0) = 1 - \Phi(0) = \frac{1}{2}$$

2. Let $X \sim N(68, 3^2)$ and $Y \sim N(66, 2^2)$ be independent. $P(X > Y)$ equals

- (a) $1 - \Phi\left(\frac{0-2}{\sqrt{3^2+2^2}}\right)$
- b $1 - \Phi\left(\frac{0-2}{3^2+2^2}\right)$
- c $1 - \Phi\left(\frac{68-66}{\sqrt{3^2+2^2}}\right)$
- d none of the above

Soln

$$X-Y \sim N(68-66, 3^2+2^2) = N(2, 13).$$

$$\begin{aligned} P(X > Y) &= P(X-Y > 0) \\ &= P\left(\frac{(X-Y)-2}{\sqrt{13}} > \frac{0-2}{\sqrt{13}}\right) \\ &= 1 - P\left(\frac{(X-Y)-2}{\sqrt{13}} < \frac{-2}{\sqrt{13}}\right) \\ &= 1 - \Phi\left(-\frac{2}{\sqrt{13}}\right) \end{aligned}$$

Sec 5.5 Chi-square distribution

see end of lecture notes for proof.

Fact If $Z \sim N(0,1)$ then

$$Z^2 \sim \text{Gamma}\left(\frac{r}{2}, \frac{1}{2}\right)$$

Recall that if $T \sim \text{Gamma}(r, \lambda)$, $M_T(t) = \left(\frac{\lambda}{\lambda-t}\right)^r$, $t < \lambda$

Hence $M_{Z^2}(t) = \left(\frac{\frac{1}{2}}{\frac{1}{2}-t}\right)^{\frac{r}{2}}, t < \frac{1}{2}$

Let $Z_1, \dots, Z_n \stackrel{\text{iid}}{\sim} N(0,1)$

$$M_{\sum Z_i^2}(t) = \left(\frac{\frac{n}{2}}{\frac{1}{2}-t}\right)^{\frac{n}{2}}, t < \frac{1}{2}$$

By uniqueness of MGF

$$\sum Z_i^2 \sim \text{Gamma}\left(\frac{n}{2}, \frac{1}{2}\right)$$

(called chi-square distribution, χ_n^2 , with n degrees of freedom)

Ex Let $X, Y \stackrel{\text{iid}}{\sim} N(0,1)$

Let $R = \sqrt{x^2 + y^2}$ be Raleigh distribution.

Then $R^2 = X^2 + Y^2 \sim \chi_2^2$ (we should test this $R^2 \sim \text{Exp}\left(\frac{1}{2}\right)$
" " $\text{Gamma}(1, \frac{1}{2})$)

Sec 5.4 The Density Convolution Formula

Ex Let $X \sim \text{Unif}\{0, 1, 2, 3, 4, 5, 6\}$

$Y \sim \text{Unif}\{0, 1, 2, 3, 4, 5, 6\}$

Let $S = X + Y$.

Find the probability mass function of S .

$$P(S=3) = P(X=0, Y=3-0) + P(X=1, Y=3-1) + P(X=2, Y=3-2) \\ + P(X=3, Y=3-3)$$

$$P(S=s) = \sum_{x=0}^s P(X=x, Y=s-x)$$

Convolution formula.

Let $X > 0, Y > 0$ be continuous RVs.

Let $S = X + Y$

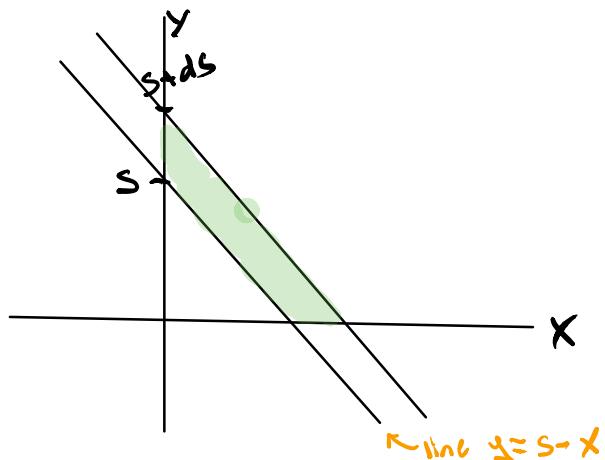
Find the density of S

$$s = x + y$$

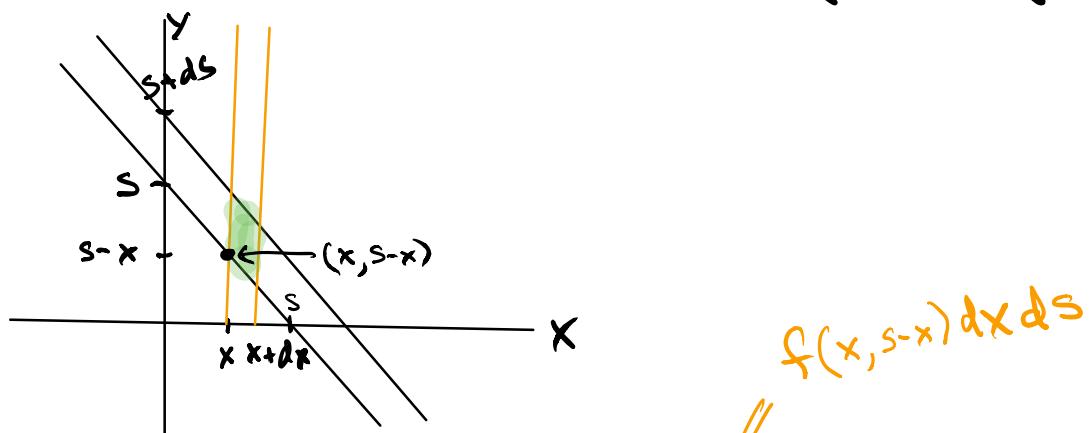
$$y = s - x$$

of interest.

$P(S \in ds)$ is the volume under $f(x, y)$ over the green region.



$P(S \in ds, X \in dx)$ is the volume under $f(x, y)$ over the green region.



$$\begin{aligned} P(S \in ds) &= \int_{x=0}^{x=s} P(S \in ds, X \in dx) \\ &\quad // \\ &\quad f(s) ds \\ &= \int_{x=0}^{x=s} f(x, s-x) dx ds \\ &\quad \boxed{x=0 \quad f(s)} \\ &\quad \boxed{x=s} \end{aligned}$$

$$\Rightarrow f(s) = \int_{x=0}^s f(x, s-x) dx$$

Convolution formula for densities.

$$x, y \stackrel{\text{iid}}{\sim} \text{expon}(\lambda) \quad S = X + Y$$

$$\begin{aligned}
 f_S(s) &= \int_0^s f(x, s-x) dx \\
 &= \int_0^s f_X(x) f_Y(s-x) dx \\
 &= \int_0^s \lambda e^{-\lambda x} \lambda e^{-\lambda(s-x)} dx \\
 &= \int_0^s \lambda^2 e^{-\lambda s} dx = \lambda^2 e^{-\lambda s} x \Big|_0^s \\
 &= \boxed{\lambda^2 e^{-\lambda s} \cdot s}
 \end{aligned}$$

variable part of
gamma(2, λ)

$\Rightarrow S \sim \text{gamma}(2, \lambda)$.

\Rightarrow Let $X \sim U_{(0,1)}$, $Y \sim U_{(0,1)}$ for 10 iid $U(0,1)$.
 The joint density $f_{X,Y}(x,y) = \binom{10}{6,1,1,1} x^6 (y-x)^1 (1-y)^1$
 for $0 < x < y < 1$.

$$\text{Let } Z = Y - X$$

- a) For a fixed Z , what is the smallest and largest value of x ?

$$Z = y - x$$

$$x = y - z \quad \text{where } y \in [0,1]$$

x must be > 0 and < 1

smallest: 0 (let $y = z$)

largest: $1 - x$ (let $y = 1$)

so convolution formula for $Z = Y - X$ is

$$f_Z(z) = \int_{x=0}^{x=1-z} f_{X,Y}(x, x+z) dx$$

\Rightarrow Let $X \sim U_{(7)}, Y \sim U_{(9)}$ for 10 iid $U(0,1)$.
 The joint density $f_{X,Y}(x,y) = \binom{10}{6,1,1,1} x^6 (y-x)(1-y)$
 for $0 < x < y < 1$.

Prove via the convolution formula that
 the density of $Z = Y - X$ is a Beta distribution
 and find the parameters.

$$\text{Let } C = \binom{10}{6,1,1,1}$$

By convolution formula,

$$\begin{aligned} f_Z(z) &= \int_0^{1-z} f_{X,Y}(x, x+z) dx \\ &= \int_0^{1-z} C x^6 (x+z-x)(1-(x+z)) dx \\ &= Cz \int_0^{1-z} ((1-z)x^6 - x^7) dx \\ &= Cz \left[\frac{(1-z)x^7}{7} - \frac{x^8}{8} \right]_{x=0}^{x=1-z} \\ &= Cz \left(\frac{(1-z)^8}{7} - \frac{(1-z)^8}{8} \right) \end{aligned}$$

Ignoring constants, $f_Z(z) \propto z(1-z)^7$, $0 < z < 1 \Rightarrow Z \sim \text{Beta}(2, 9)$

Appn ix

If $Z \sim N(0, 1)$, then $Z^2 \sim \text{Gamma}(\frac{r}{2}, \frac{1}{2})$

Proof /

$\lambda > 0$, $r >$ integer r ,

gamma (r, λ) density

$$f(t) = \frac{\lambda^r}{\Gamma(r)} t^{r-1} e^{-\lambda t}, \quad t > 0$$

let $Z = \text{std normal}$
change of variable rule.

$$X = Z^2$$

$$\text{Find } f_X(x) = \frac{1}{\sqrt{2\pi}} X^{-\frac{1}{2}} e^{-\frac{1}{2}x}, \quad x > 0$$

$$= \frac{1}{\sqrt{\pi}} X^{\frac{1}{2}-1} e^{-\frac{1}{2}x}$$

$$\times \Gamma(\frac{1}{2})$$

$$\Rightarrow X \sim \text{gamma}(\frac{r}{2}, \frac{1}{2})$$

□