

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 $\xi_1, \xi_2 \stackrel{iid}{\sim} N(0, 1)$
- ① $\xi_1^2 + \xi_2^2 = T \sim \text{Exp}(\frac{1}{2})$.
 - ② $\Phi(z) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^z e^{-\frac{1}{2}t^2} dt, E(z) = 0, \text{SD}(z) = 1$
- c) If $z \sim N(0, 1)$, $X = \mu + \sigma z \sim N(\mu, \sigma^2)$
 with density $f(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2}(\frac{x-\mu}{\sigma})^2}$
- d) $\xi_1 + \xi_2 \sim N(0, 2)$

Today

sec 5.3

① Review MGF

① A linear combination of normal is normal

② Chi square distribution

sec 5.4

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

MGF Review (lecture 25)

Recall the MGF of a RV X is $M(t) = E(e^{tX})$ for some interval of t containing 0.

Common MGF

$$X \sim N(0,1), M(t) = e^{\frac{t^2}{2}} \text{ for all } t$$

$$T \sim \text{Gamma}(r, \lambda), M(t) = \left(\frac{\lambda}{\lambda-t}\right)^r \text{ for } t < \lambda$$

Properties of MGF

$$\text{if } X, Y \text{ are independent } M_{X+Y}(t) = M_X(t)M_Y(t)$$

$$\text{if } M_X(t) = M_Y(t) \text{ then } X = Y$$

$$M_{b+aX}(t) = e^{bt} M(aX)$$

ex Use these properties to find the MGF of $X \sim N(\mu, \sigma^2)$

Soln

For $Z \sim N(0,1)$

$$X = \mu + \sigma Z \quad \text{and} \quad M_Z(t) = e^{\frac{t^2}{2}}$$

$$\text{so } M_X(t) = e^{\mu t} \underbrace{e^{\frac{\sigma^2 t^2}{2}}}_{M_Z(\sigma t)} = e^{\frac{(\mu + \sigma t)^2}{2}}$$

Q Use MGF to prove that the sum of two independent std normals is $N(0, 2)$.

Soln $z_1, z_2 \stackrel{iid}{\sim} N(0, 1)$

$$\begin{aligned} M_{z_1+z_2}(t) &= M_{z_1}(t)M_{z_2}(t) \\ &= e^{\frac{t^2}{2}} \cdot e^{\frac{t^2}{2}} = e^{t^2} \end{aligned}$$

$$e^{t^2}$$

by uniqueness of MGF

$$z_1+z_2 \sim N(0, 2)$$

Thm Let $x_1 \sim N(\mu_1, \sigma_1^2)$
 $x_2 \sim N(\mu_2, \sigma_2^2)$ } indep.

then $a x_1 + b x_2 \sim N(a\mu_1 + b\mu_2, a^2 \sigma_1^2 + b^2 \sigma_2^2)$

Pf at end of lecture.

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

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 $X - ZY \sim N(0, 3)$

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

normalize $\xrightarrow{\text{normalize}} = P\left(\frac{(X - ZY) - 0}{\sqrt{3}} > \frac{0 - 0}{\sqrt{3}}\right)$

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

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

$$\begin{aligned}
 &\xrightarrow{\text{soln}} X-Y \sim N(68-66, 3^2+2^2) = N(2, 13), \\
 P(X > Y) &= P(X-Y > 0) \\
 &= P\left(\frac{(X-Y)-2}{\sqrt{13}} > \frac{0-2}{\sqrt{13}}\right) \\
 &= 1 - P\left(\frac{(X-Y)-2}{\sqrt{13}} < \frac{-2}{\sqrt{13}}\right) \\
 &= 1 - \Phi\left(-\frac{2}{\sqrt{13}}\right)
 \end{aligned}$$

Sec 5.3 Chi-square distribution

see end of lecture notes for proof.

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

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

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

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

Let $Z_1, \dots, Z_n \stackrel{i.i.d.}{\sim} N(0,1)$

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

By uniqueness of MGF

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

↑ called chi-square distribution with n degrees of freedom.

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 $X \geq 0, Y \geq 0$ be continuous RVs.

Let $S = X + Y$

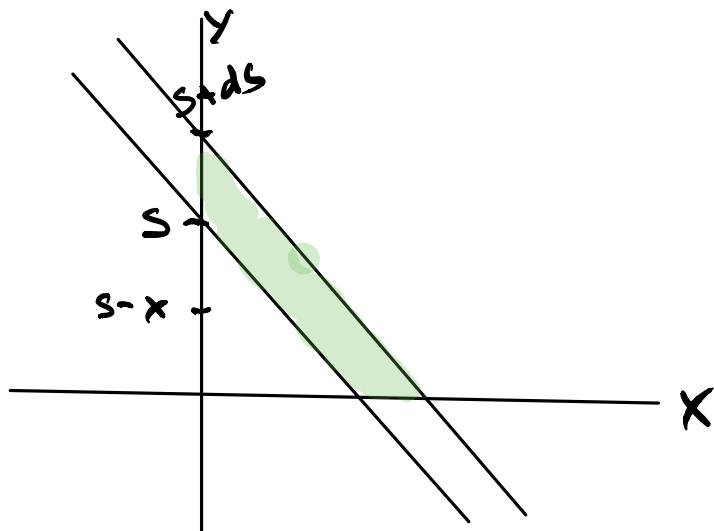
Find the density of S

$$s = x + y$$

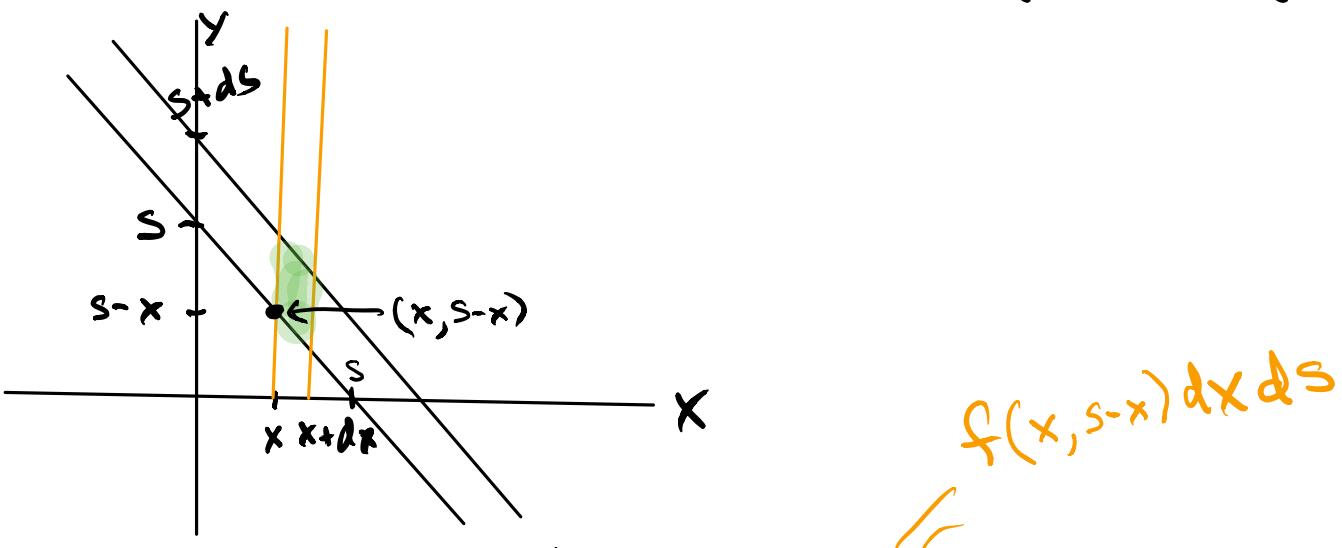
$$y = s - x$$

↗ y intercept.

$P(S \in ds)$ 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.



$$P(S \in ds) = \int_{x=0}^{x=s} P(S \in ds, X \in dx) f(x, s-x) dx ds$$

$$\text{f}(s)ds$$

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

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

Convolution formula for densities.

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

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

variable part of
gamma(2, λ)

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

Appendix

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

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

Pf) $M_{X_1}(t) = e^{\mu_1 t} e^{\frac{\sigma_1^2 t^2}{2}}$

$$M_{aX_1}(at) = M_{X_1}(at) = e^{\mu_1 at} e^{\frac{\sigma_1^2 a^2 t^2}{2}}$$

$$\text{then } M_{aX_1 + bX_2}(t) = M_{aX_1}(t) M_{bX_2}(t)$$

$$= e^{\mu_1 at} e^{\frac{\sigma_1^2 a^2 t^2}{2}} \cdot e^{\mu_2 bt} e^{\frac{\sigma_2^2 b^2 t^2}{2}}$$

$$= e^{(a\mu_1 + b\mu_2)t} e^{\frac{(\sigma_1^2 a^2 + \sigma_2^2 b^2)t^2}{2}}$$

by uniqueness of MGF

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

□

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

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

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

$$X = Z^2$$

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

$$= \frac{\left(\frac{1}{2}\right)^{\frac{1}{2}}}{\sqrt{\pi}} x^{\frac{1}{2}-1} e^{-\frac{1}{2}x}$$
$$\Gamma\left(\frac{1}{2}\right)$$

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

□