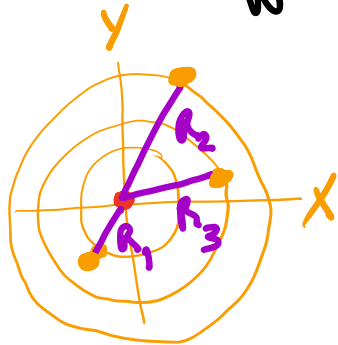


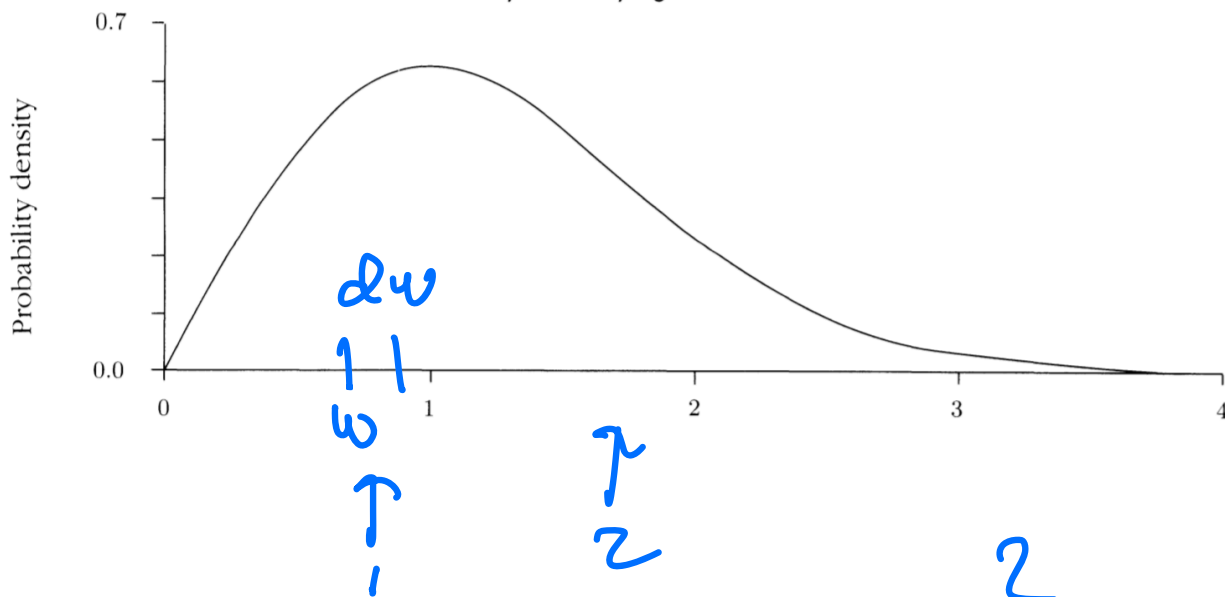
warmup

Suppose 3 shots are fired at a target. Assume for each shot, both X and Y are standard normals. Let W be the closest distance among 3 shots to the bullseye. Find $f_W(w)$

Picture



Hint: $W = \min(R_1, R_2, R_3)$
where $R_1, R_2, R_3 \stackrel{iid}{\sim} Ray$

FIGURE 3. Density of the Rayleigh distribution of R .

$$f_W(w) = \binom{3}{1,2} f(w) (1 - F(w))^2$$

$$= \binom{3}{1,2} w e^{-\frac{1}{2}w^2} \left(e^{-\frac{1}{2}w^2} \right)^2$$

Last time

sec 5.3 Rayleigh distribution

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

and CDF $F_R(r) = 1 - e^{-\frac{1}{2}r^2}$, $r > 0$.

We saw last time that $R = \sqrt{X^2 + Y^2}$
where $X, Y \stackrel{\text{iid}}{\sim} N(0, 1)$.

We proved that the constant
of the density of standard normal is
 $c = \frac{1}{\sqrt{2\pi}}$ (i.e. $\phi(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}}$ is the density
of $Z \sim N(0, 1)$).

Today

- ① sec 5.3 Sum of independent normals
- ② sec 5.3 Chi square distribution

① sec 5.3 Sum of independent normals

Definition of the Normal (μ, σ^2) distribution:

We know the density of the std normal

$$\phi(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}}$$

Make a change of scale $X = \mu + \sigma Z$. By definition $X \sim N(\mu, \sigma^2)$.
By the change of variable rule you can show

$$f_X(x) = \frac{1}{\sigma} \phi\left(\frac{x-\mu}{\sigma}\right) = \frac{1}{\sigma} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}$$

MGF Review (lecture 24)

Recall the MGF of a RV X is $M_X(t) = E(e^{Xt})$.

MGF of std normal

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

Properties of MGF

If X, Y are independent then $M_{X+Y}(t) = M_X(t)M_Y(t)$

for t in an interval containing zero

If $M_X(t) = M_Y(t)$ then $X \stackrel{d}{=} Y$

same distribution

$$M_{\mu + \sigma Z}(t) = e^{\mu t} M_Z(\sigma t)$$

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

Hint let $Z \sim N(0,1)$, $\mu \in \mathbb{R}$, $\sigma > 0$
and $X = \mu + \sigma Z$

$$\begin{aligned} M_X(t) &= M_{\mu + \sigma Z}(t) = e^{\mu t} M_Z(\sigma t) \\ &= e^{\mu t} e^{-\frac{1}{2}(\sigma t)^2} \\ &= e^{\mu t} e^{-\frac{\sigma^2 t^2}{2}} \quad \text{for all } t \in \mathbb{R} \end{aligned}$$

if $X \sim N(0, \overset{\sigma^2}{2})$

$$M_X(t) = e^{-\frac{t^2}{2}}$$

Q1 Use MGF to prove that the sum of two independent std normals is $N(0, 2)$.

hint

$$X \sim N(\mu, \sigma^2) \Rightarrow M_X(t) = e^{\mu t} e^{\frac{\sigma^2 t^2}{2}}$$

$$\begin{aligned} M_{Z_1 + Z_2}(t) &= M_{Z_1}(t) M_{Z_2}(t) \\ &= e^{t^2/2} \cdot e^{t^2/2} \\ &= e^{t^2} \text{ for all } t \end{aligned}$$

\hookrightarrow MGF of $N(0, 2)$

Hint $X \sim N(0, 2)$ $\swarrow \sigma^2$

$$M_X(t) = e^{t^2}$$

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

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

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

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

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

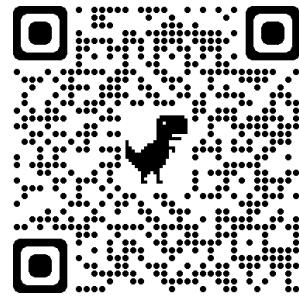
$= e^{\mu_1 at} e^{\frac{\sigma_1^2 a^2 t^2}{2}} \cdot e^{\mu_2 bt} e^{\frac{\sigma_2^2 b^2 t^2}{2}}$

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

by uniqueness of MGF

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

□

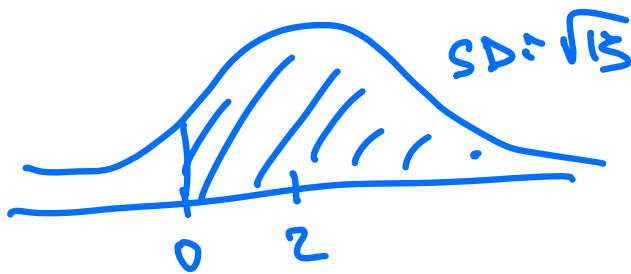


2. Let $X \sim N(68, 3^2)$ and $Y \sim N(66, 2^2)$ be independent. $P(X > Y)$ equals

- 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

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

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



area is

$$1 - \Phi\left(\frac{0-2}{\sqrt{3^2+2^2}}\right)$$

② Sec 5.3 Chi-square distribution

Fact — *see end of lecture notes*
if $Z \sim N(0,1)$ then

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

Recall that if $T \sim \text{Gamma}(r, \lambda)$,

$$M_T(t) = \left(\frac{\lambda}{\lambda - t}\right)^r, \quad t < \lambda$$

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

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

$$M_{Z_1^2 + \dots + Z_n^2} = \left(\frac{\frac{1}{2}}{\frac{1}{2} - t}\right)^{\frac{1}{2} \cdot n}, \quad t < \frac{1}{2}$$

By uniqueness of MGF

$$\underbrace{Z_1^2 + \dots + Z_n^2} \sim \text{Gamma}\left(r = \frac{n}{2}, \lambda = \frac{1}{2}\right)$$

*called chi squared distribution (χ^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 Rayleigh distribution.

Then $R^2 = X^2 + Y^2 \sim \chi_2^2 \leftarrow \text{\# degrees of freedom}$

extra (Rayleigh)

If $R_1 \sim \text{Ray}$ (density $f_R(r) = r e^{-\frac{1}{2}r^2}$)
find the density of $W = \frac{1}{\sqrt{3}} R_1$

$$f_W(w) = \frac{1}{\frac{1}{\sqrt{3}}} \cdot r e^{-\frac{1}{2}r^2} \Big|_{r=\sqrt{3}w}$$
$$= \sqrt{3} \cdot \sqrt{3}w e^{-\frac{3}{2}w^2} = 3w e^{-\frac{3}{2}w^2}$$

So $W = \frac{1}{\sqrt{3}} \text{Ray}$ is also the minimum of R_1, R_2, R_3
i.e. $\min(R_1, R_2, R_3) \sim \frac{1}{\sqrt{3}} \text{Ray}$

or

Let $R_1, R_2, R_3, R_4 \stackrel{\text{i.i.d.}}{\sim} \text{Ray}$

Let $W = \min(R_1, R_2, R_3) \sim \frac{1}{\sqrt{3}} \text{Ray}$

Find $P(R_4 < W)$

Hint: Find $P(R_4^2 < W^2)$

Recall
 $\text{Exp}(\lambda) = \text{Exp}\left(\frac{\lambda}{2}\right)$

$$P(R_4 < W) = P(R_4^2 < W^2)$$

$$R_4^2 \sim \text{Exp}(\lambda) \text{ and } W^2 \sim \frac{1}{3} \text{Ray}^2 = \frac{1}{3} \text{Exp}\left(\frac{1}{2}\right) = \text{Exp}\left(\frac{3}{2}\right)$$

$$P\left(\text{Exp}\left(\frac{1}{2}\right) < \text{Exp}\left(\frac{3}{2}\right)\right) = \frac{\frac{1}{2}}{\frac{1}{2} + \frac{3}{2}} = \boxed{\frac{1}{4}}$$

Competing exponentials

Appendix

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

Proof/

$\lambda > 0$, pos integer r ,

gamma (r, λ) density

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

let $Z = \text{std normal}$

$X = Z^2$ change of variable rule.

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

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

$$\frac{\sqrt{\pi}}{\Gamma(\frac{1}{2})}$$

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

□