

Last timeSec 5.3 independent normal variables

- we used MGF to prove

① a linear combination of **independent** normals is normal

② If  $X_1, \dots, X_n \stackrel{iid}{\sim} N(0, 1)$  then

$$X_1^2 + X_2^2 + \dots + X_n^2 \sim \text{Gamma}(r = \frac{n}{2}, \lambda = 1)$$

~ central chi-square distribution  
with  $n$  degrees of freedom  
written  $\chi_n^2$ .

Sec 5.4 Convolution formula for density of sum

Let  $X, Y$  be RVs and  $S = X + Y$ ,

if  $X > 0, Y > 0$   $\int_{x=0}^{x=s} f(x, s-x) dx = \int_x^s f(x) f(s-x) dx$   
**If  $X, Y$  independent**

more generally

$$f_S(s) = \int_{x=-\infty}^{x=\infty} f(x, s-x) dx = \int_x^s f_x(x) f_y(s-x) dx$$

Today

① Go over student comments from concept test last time.

Finish Sec 5.4

② Example:  $X, Y \stackrel{iid}{\sim} U(0, 1)$ .  $S = X + Y$ . Find  $f_S(s)$ .

③ Convolution formula of ratio  $Y/X$

④ Proof that  $f_p(r) = \frac{\Gamma(r+s)}{\Gamma(r)\Gamma(s)} p^{r-1} (1-p)^{s-1}$ ,  $0 < p < 1$  is a density for  $r > 0, s > 0$ .

## 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

Discuss with neighbor your solution for 1 min,

$$\begin{aligned}
 &\text{Solu} \\
 &\bar{Y} \sim N(0, 1) \Rightarrow ZY \sim N(0, 4) \\
 &\text{and } X - ZY \sim N(0, 1^2 + 2^2) = N(0, 5) \\
 &P(X > ZY) = P(X - ZY > 0) \\
 &\text{normalize} = P\left(\frac{(X - ZY) - 0}{\sqrt{5}} > \frac{0 - 0}{\sqrt{5}}\right) \\
 &= P(Z > 0) = 1 - \Phi(0) = \frac{1}{2}
 \end{aligned}$$

11/2/2018 c	a	cuz the sum $2y-x$ is normal( $0,3$ ) which is still symmetric about 0
11/2/2018 9:37:09 a	a	evaluating the integral of the joint density over the plane gives $1/2$ which is equal to $1-\Phi(0) = 1/2$
11/2/2018 a	a	$\gamma$ $y = \frac{1}{2}x$ Wait, the image is symmetry on the plane!
11/2/2018 9:37:18 a	a	$\text{I think it's } a,$ as $X - 2Y > 0$ has $X - 2Y$ has $N(0,5).$ Therefore, cdf is $\Phi(0)$ for $P(X-2Y < 0).$ We have $P(X>2Y).$ Transfer the $Y$ to the left to get $P(X-2Y>0).$ $X-2Y \sim N(0,5)$ (just square and add the SDs). Now we can find the probability. Since $\mu$ is 0 and our $x$ is 0, $x-\mu/\sigma$ is 0 so we get $1-\Phi(0)$
11/2/2018 9:38:40 a	a	
11/2/2018 9:39:26 a	a	

thanks!

I have a question to the answer.  
Since  $ax_1 + bx_2 \sim N(a\mu_1 + b\mu_2, a^2 \sigma^2 + b^2 \sigma^2)$  why  $x-2y$  does not follow  $N(0, 1^2 * 1 + (-2)^2 * 1 = 5)$  instead of the answer  $N(0,3)?$

11/4/2018 a

11/2/2018 a

$Z=X-2Y \sim \text{Normal}(0,5),$   
 $P(Z>0)=1-\Phi(0)$

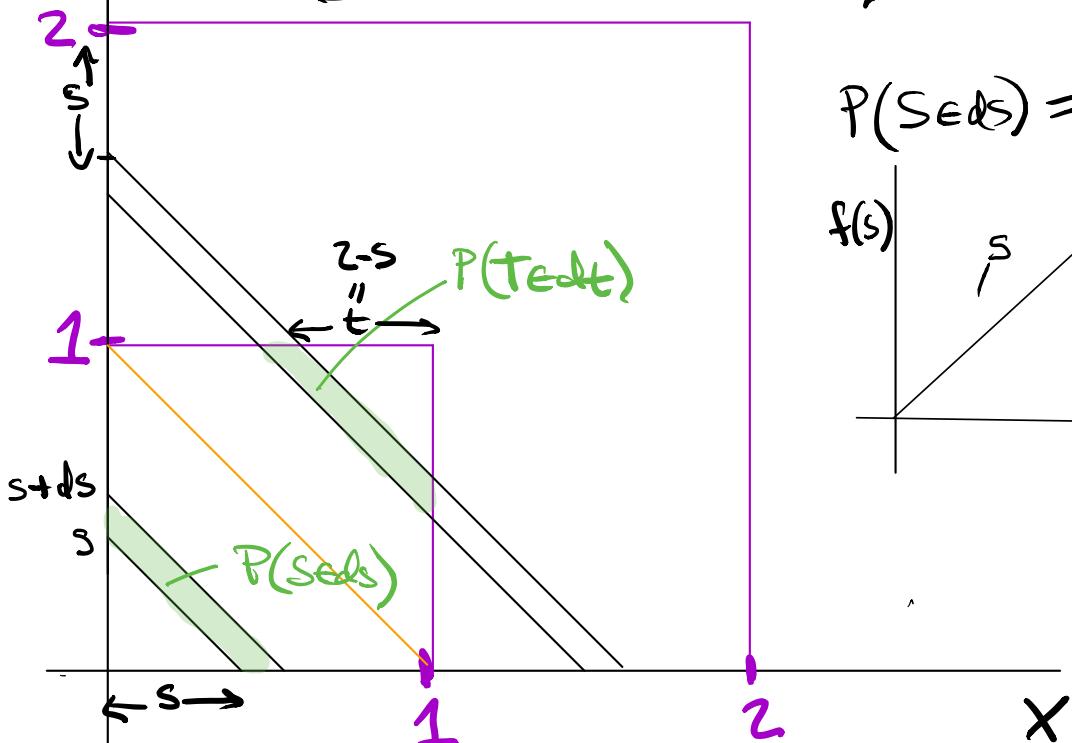
Screen Shot 2018-11-04 at 9:31.12 PM

half of all  $Y$  values will be between  $1/2$  and  $1$ , which means  $x$  cannot be greater for half of the time. thus the probability must be less than  $1/2$ . We can compare  $x \sim N(0,1)$  and  $y \sim N(0,2)$  to get our answer

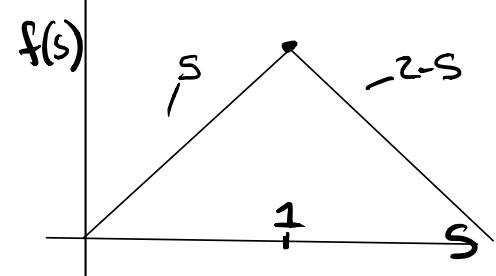
(2) Example convolution formula of sum:

ex.  $X, Y \stackrel{iid}{\sim} \text{Unif}(0,1)$ ,  $S = X + Y$

Find  $f_S(s)$ .



$$P(S=s, T=t) = f(s) ds$$



For  $0 < s < 1$

$$f(s) = \int_{x=0}^{s=x} f_x(x) f_y(s-x) dx = \int_{x=0}^{s=x} 1 \cdot 1 dx = s$$

For  $1 < s < 2$

Note if  $X, Y \stackrel{iid}{\sim} U(0,1)$  then  $1-X, 1-Y \stackrel{iid}{\sim} U(0,1)$

$$\text{Let } T = 2-S = (1-X) + (1-Y)$$

then  $0 < t < 1$

By convolution formula for T,  $\frac{f_T(t)}{T} = t$ ,  $0 < t < 1$

$$\text{so } f_S(s) = \begin{cases} s & \text{for } 0 < s < 1 \\ 2-s & \text{for } 1 < s < 2 \end{cases}$$

Knowing  $f_T(t) = t$  we find  $f_T(s) = 2-s$   
by change of variable formula

(3)

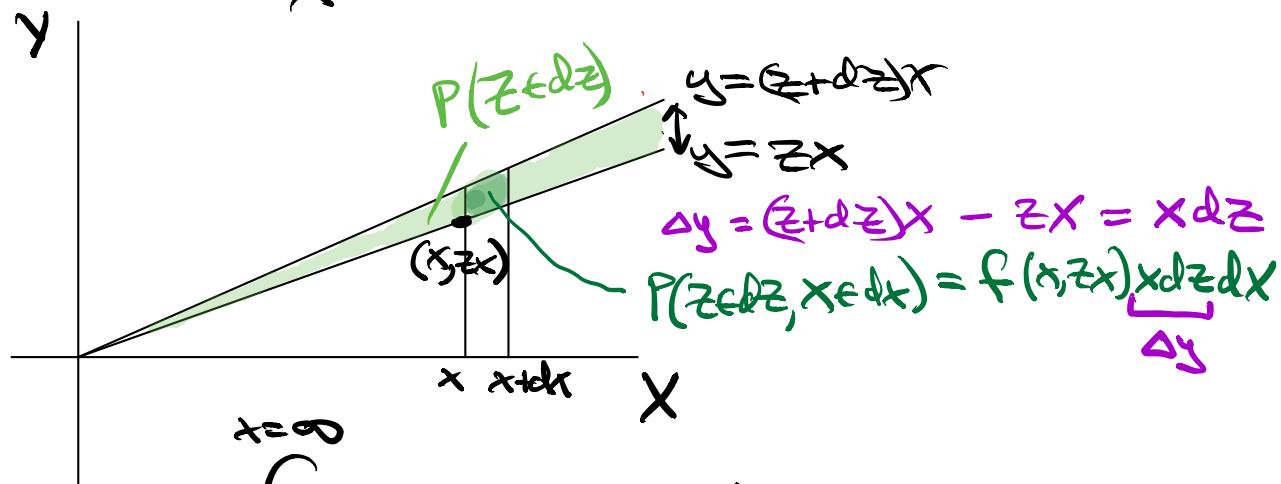
### Convolution formula for density of ratio $y/x$

$$x > 0, y > 0$$

$$\text{let } z = \frac{y}{x}.$$

Find  $f_z(z)$ .

Picture  $z = \frac{y}{x} \rightarrow y = zx$  slope.



$$P(z \in dz) = \int_{-\infty}^{+\infty} P(z \in dz, x \in dx)$$

$$= f(z)dz$$

$$= \int_{-\infty}^{+\infty} f(x, zx) x dz dx$$

$$\Rightarrow f(z) = \int_{-\infty}^{+\infty} \int_{-\infty}^{+\infty} f(x, zx) x dx dz = \int_{-\infty}^{+\infty} f(x) f_y(zx) x dx$$

**Convolution formula.**

Recall gamma density

$$X \sim \text{Gamma}(r, \lambda)$$

$$f_X(x) = \frac{\lambda^r}{\Gamma(r)} x^{r-1} e^{-\lambda x}$$

↑ variable part

$$\text{so } \int_0^\infty x^{r-1} e^{-\lambda x} dx = \frac{\Gamma(r)}{\lambda^r}$$

Q8 Let  $X, Y \stackrel{\text{iid}}{\sim} \text{Exp}(1)$ .  $Z = \frac{Y}{X}$   
Find  $f_Z(z)$ .

Soln  $f_X(x) = e^{-x}, x \geq 0.$

$$f_Z(z) = \int_0^\infty e^{-x} e^{-zx} x dx$$

$$= \int_{x=0}^\infty x e^{-(1+z)x} dx$$

gamma( $r=z, \lambda=1+z$ )

$$= \frac{\Gamma(z)}{(1+z)^z} = \boxed{\frac{1}{(1+z)^z}} \quad \text{for } z > 0.$$

# Concrt test

## Stat 134

Monday November 5 2018

1. Let  $U \sim U(0, 1)$  and  $V \sim U(0, 1)$  be independent. The density of  $Y = U/V$  for  $0 < y < 1$  is:

a  $1/(2y)$

b  $1/2$

c  $1/(2y^2)$

d none of the above

Soln

$$Y = \frac{U}{V}$$

$v=1$  since  
need  $v < 1$   
need  $uv < 1$

$$f_Y(y) = \int_{v=0}^1 f_V(v) f_U(yv) v dv$$

$$= \int_0^1 1 \cdot 1 \cdot v dv = \frac{v^2}{2} \Big|_0^1 = \boxed{\frac{1}{2} \text{ for } 0 < y < 1}$$

2. Let  $U \sim U(0, 1)$  and  $V \sim U(0, 1)$  be independent. The density of  $Y = U/V$  for  $y > 1$  is:

**a**  $1/(2y)$

**b**  $1/2$

**c**  $1/(2y^2)$

**d** none of the above

true if  $v < \frac{1}{y}$

for  $y > 1$

$$Y = \frac{U}{V} \quad v = \frac{1}{y}$$

$$f_Y(y) = \int_{v=0}^{\infty} f_V(v) f_U(yv) v dv$$

need  $v < 1$   
need  $gv < 1 \Rightarrow v < \frac{1}{g}$

$$= \int_0^{1/y} 1 \cdot 1 v dv = \frac{1}{2} \left[ v^2 \right]_0^{1/y} = \boxed{\frac{1}{2y^2} \text{ for } y > 1}$$

(4) In sec 4.6 we defined the Beta distribution  $P \sim \text{Beta}(r, s)$  for  $r > 0, s > 0$  to have density

$$f_P(p) = \frac{\Gamma(r+s)}{\Gamma(r)\Gamma(s)} p^{r-1} (1-p)^{s-1}, \quad 0 < p < 1$$

**WE still need to show that this is a density!**

we know:

① Gamma  $r > 0, \lambda > 0$

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

$$\text{where } \Gamma(r) = \int_0^\infty x^{r-1} e^{-x} dx, \quad r > 0$$

② Convolution formula for sum,  $Z = X + Y$  where  $X > 0, Y > 0$  independent.

$$f_Z(z) = \int_{x=0}^{x=z} f_X(x) f_Y(z-x) dx \quad \leftarrow \text{This is a density}$$

$$\left. \begin{array}{l} \text{Let } X \sim \text{Gamma } (r, \lambda) \\ \text{Let } Y \sim \text{Gamma } (s, \lambda) \end{array} \right\} \text{indep.}$$

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

By the convolution formula for  $Z = X + Y$   
we know  $f_Z(z)$  below is a density:

$$\begin{aligned}
f_Z(z) &= \int_0^z f_X(x) f_Y(z-x) dx, \quad z > 0 \\
&= \int_0^z \frac{\lambda^{r-1} e^{-\lambda x}}{\Gamma(r)} \cdot \frac{\lambda^{s-1} e^{-\lambda(z-x)}}{\Gamma(s)} dx \\
&= \int_0^z \frac{\lambda^{r+s}}{\Gamma(r)\Gamma(s)} x^{r-1} (z-x)^{s-1} e^{-\lambda z} dx \\
&= \frac{\lambda^{r+s}}{\Gamma(r)\Gamma(s)} e^{-\lambda z} \int_0^z x^{r-1} (z-x)^{s-1} dx
\end{aligned}$$

Let  $P = \frac{x}{z}$  to change limit of integration  
to 1.

$$x = Pz, dx = z dP$$

$$z P^{r-1} (1-P)^{s-1}$$

$$f_Z(z) = \frac{\lambda^{r+s}}{\Gamma(r)\Gamma(s)} e^{-\lambda z} \int_{P=0}^{P=1} (Pz)^{r-1} (z - Pz)^{s-1} z dP$$

$$f_z(z) = \frac{\lambda}{\Gamma(r)\Gamma(s)} \left( \int_0^1 p^{r-1} (1-p)^{s-1} dp \right) \cdot z^{r+s-1 - \lambda z} e^{-z}$$

constant part  
no  $z$ .

variable part  
( $z$ )

gamma  
( $r+s, \lambda$ )

$$= \frac{\lambda^{r+s}}{\Gamma(r)\Gamma(s)} \left( \int_0^1 p^{r-1} (1-p)^{s-1} dp \right) \frac{\Gamma(r+s)}{r+s} \cdot \frac{\lambda^{r+s}}{\Gamma(r+s)} z^{r+s-1 - \lambda z} e^{-z}$$

$$f_z(z) = \left[ \frac{\Gamma(r+s)}{\Gamma(r)\Gamma(s)} \int_0^1 p^{r-1} (1-p)^{s-1} dp \right] \frac{\lambda^{r+s}}{\Gamma(r+s)} z^{r+s-1 - \lambda z} e^{-z}$$

density of  
gamma ( $r+s, \lambda$ )

must be 1  
since  $f_z(z)$   
is a density,  
This proves that

$\frac{\Gamma(r+s)}{\Gamma(r)\Gamma(s)} p^{r-1} (1-p)^{s-1}$  is a density.

□