

Transformation!

- MGFs vs. Convolution
- Bivariate change of variables

Last time we considered $X \sim \text{Poisson}(\lambda_1)$
 $Y \sim \text{Poisson}(\lambda_2)$

For $X \perp Y$ indep. we

found that for $W = X + Y$, $W \sim \text{Poisson}(\lambda_1 + \lambda_2)$

- we had to know the binomial theorem.

Let's use MGFs to find the dist for w .

Recall $M_X(t) = \exp(\lambda_1(e^t - 1))$
and

$$M_Y(t) = \exp(\lambda_2(e^t - 1))$$

Independence implies $M_W(t) = M_X(t) M_Y(t)$

$$= \exp\left[\underbrace{(\lambda_1 + \lambda_2)}_{\lambda_1 + \lambda_2}(e^t - 1)\right]$$

$$\Rightarrow W \sim \text{Poisson}(\lambda_1 + \lambda_2)$$

a) Consider $X_1 \sim \text{Bernoulli}(p)$

$$\Rightarrow X_1 = \begin{cases} 0 & 1-p \\ 1 & p \end{cases}$$

Let X_1 and X_2 be independent. $X_2 \sim \text{Bernoulli}(p)$

Find $W = X_1 + X_2$.

$$\begin{aligned} \Rightarrow M_1(t) &= (1-p)e^{0t} + pe^t \\ &= 1-p + pe^t \end{aligned}$$

$$M_W(t) = M_1(t) M_2(t) = (1-p + pe^t)^2$$

Binomial ($n=2, p$)

$$\text{For Binomial } M_X(t) = (1-p + pe^t)^n$$

b) Suppose $X_1 \sim \text{Binomial}(n_1, p)$ X_1, X_2 are independent
 $X_2 \sim \text{Binomial}(n_2, p)$

Let $W = X_1 + X_2$. What is the dist. of W ?

$$M_W(t) = M_{X_1}(t) M_{X_2}(t) =$$

$$= (1-p + pe^t)^{n_1} (1-p + pe^t)^{n_2} =$$

$$= (1-p + pe^t)^{\underline{n_1 + n_2}} \equiv \text{Binomial}(n = n_1 + n_2, p)$$

c) Suppose $X_i \sim N(\mu_i, \sigma_i^2)$ and X_i are indep.

$W = X_1 + X_2$. Find $f_W(w)$. $M_X(t) = \exp(-\mu_i t + \frac{\sigma_i^2 t^2}{2})$

$$M_W(t) = M_1(t) M_2(t) = \exp\left[-\mu_1 t + \frac{\sigma_1^2 t^2}{2}\right] \exp\left[-\mu_2 t + \frac{\sigma_2^2 t^2}{2}\right]$$

$$= \exp\left[-(\mu_1 + \mu_2)t + (\sigma_1^2 + \sigma_2^2) \frac{t^2}{2}\right] \equiv N(\mu_1 + \mu_2, \sigma_1^2 + \sigma_2^2)$$

Let X_1 and X_2 be independent χ^2 r.v.s

$$X_1 \sim \chi^2(m), X_2 \sim \chi^2(n)$$

χ^2 is a special case of Gamma dist

$$\chi^2(m) \equiv \text{Gamma}(\alpha = \frac{m}{2}, \lambda = \frac{1}{2})$$

Suppose ~~$W = X_1 + X_2$~~ $W = X_1 + X_2$. Find $f_W(w)$

We should expect $\chi^2(m+n)$,

a) Convolution approach $W = X_1 + X_2 \Rightarrow X_2 = \underline{W - X_1}$

$$f_1(x_1) = \frac{1}{\Gamma(\frac{m}{2}) 2^{\frac{m}{2}}} x_1^{\frac{m}{2}-1} e^{-\frac{x_1}{2}}, \quad \underline{x_1 > 0}$$

↑
Gamma function

$$f_2(x_2) = \frac{1}{\Gamma(\frac{n}{2}) 2^{\frac{n}{2}}} x_2^{\frac{n}{2}-1} e^{-\frac{x_2}{2}}, \quad \underline{x_2 > 0}$$

$W - x_1 > 0 \Rightarrow w > x_1$

$$f_W(w) = \int f_1(x_1) f_2(\underline{w - x_1}) dx_1$$

$$= \int_0^w \frac{1}{\Gamma(\frac{m}{2}) 2^{\frac{m}{2}}} x_1^{\frac{m}{2}-1} e^{-\frac{x_1}{2}} \frac{1}{\Gamma(\frac{n}{2}) 2^{\frac{n}{2}}} (w - x_1)^{\frac{n}{2}-1} e^{-\frac{(w-x_1)}{2}} dx_1$$

↑
integrand

* Write the ~~integral~~ as a known pdf
w/ support b/w 0 and w , we are done.

- Use Beta dist !!

$$\int_0^1 \frac{\Gamma(\frac{m+n}{2})}{\Gamma(\frac{m}{2}) \Gamma(\frac{n}{2})} y^{\frac{m}{2}-1} (1-y)^{\frac{n}{2}-1} dy = 1$$

1) Multiply by $\frac{\Gamma(\frac{m+n}{2})}{\Gamma(\frac{m}{2})\Gamma(\frac{n}{2})}$

2) $X = wy \Rightarrow Y = \frac{x}{w} \quad x \rightarrow 0 \Rightarrow y \rightarrow 0$

$dx = w dy \quad x \rightarrow w \Rightarrow y \rightarrow 1$

$$\Rightarrow \int_0^1 \frac{1}{2^{\frac{m+n}{2}} \Gamma(\frac{m+n}{2})} \int_0^1 \frac{\Gamma(\frac{m+n}{2})}{\Gamma(\frac{m}{2})\Gamma(\frac{n}{2})} (wy)^{\frac{m}{2}-1} e^{-\frac{w}{2}} (w-wy)^{\frac{n}{2}-1} w dy$$

$$= \frac{w^{\frac{m}{2}-1} w^{\frac{n}{2}-1} w e^{-\frac{w}{2}}}{\Gamma(\frac{m+n}{2}) 2^{\frac{m+n}{2}}} \int_0^1 \frac{\Gamma(\frac{m+n}{2})}{\Gamma(\frac{m}{2})\Gamma(\frac{n}{2})} y^{\frac{m}{2}-1} (1-y)^{\frac{n}{2}-1} dy$$

$$= \frac{w^{\frac{m+n}{2}-1} e^{-\frac{w}{2}}}{\Gamma(\frac{m+n}{2}) 2^{\frac{m+n}{2}}} \quad \text{by def. of Beta pdf.}$$

$$\underline{\underline{= \chi^2(m+n)}}$$

b) MGF's $M_1(t) = \left(\frac{1}{1-2t}\right)^{m/2}, \quad M_2(t) = \left(\frac{1}{1-2t}\right)^{n/2}$

$$M_w(t) = M_1(t) M_2(t) = \left(\frac{1}{1-2t}\right)^{\frac{m+n}{2}}$$

Suppose X & Y are independent discrete r.v.'s

$$P_X(1) = .2, P_X(2) = .4, P_X(3) = .3, P_X(4) = .1$$

$$P_Y(1) = .3, P_Y(3) = .5, P_Y(5) = .2$$

Let $W = X + Y$. Find $P_W(w)$ using MGFs.

$$M_W(t) = M_X(t) M_Y(t)$$

$$= (.2e^t + .4e^{2t} + .3e^{3t} + .1e^{4t}) \cdot$$

$$(.3e^t + .5e^{3t} + .2e^{5t})$$

$$= .06e^{2t} + .12e^{3t} + .19e^{4t} + .23e^{5t} + .19e^{6t} + .13e^{7t} \\ + .06e^{8t} + .02e^{9t}$$

<u>W</u>	<u>$P(W=w)$</u>
2	.06
3	.12
4	.19
5	.23
6	.19
7	.13
8	.06
9	.02

X	0	1	2
1	.15	.10	0
2	.25	.30	.20

Find the joint pmf
of $u = X+Y$ and $V = XY$.

X	$Y=0$
1	.15
	$u=1$
	$V=0$

$Y=1$
.10
$u=2$
$V=1$

$Y=2$
0
$u=1$
$V=2$

2	.25
	$u=2$
	$V=0$

.30
$u=3$
$V=2$

.20
$u=4$
$V=4$

u	0	1	2	4
1	.15	0	0	0
2	.25	.10	0	0
3	0	0	.30	0
4	0	0	0	.20

Bivariate Change of Variable,

Let X_1 and X_2 have joint pdf $f(x_1, x_2)$

$$\text{val} \text{ suppo.} \quad S = \{(x_1, x_2) : f(x_1, x_2) > 0\}$$

$$\text{Let } Y_1 = u_1(x_1, x_2) \quad ; \quad Y_2 = u_2(x_1, x_2)$$

u_1 and u_2 to one-to-one transformations.

$$\Rightarrow X_1 = w_1(y_1, y_2) \quad ; \quad X_2 = w_2(y_1, y_2)$$

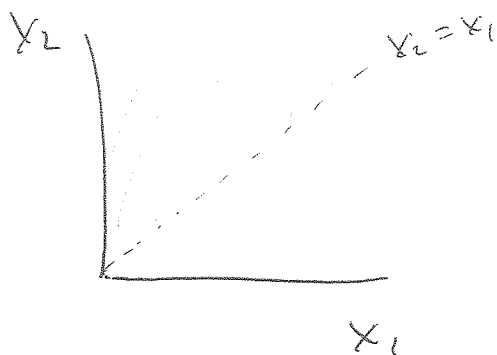
$$J = \begin{vmatrix} \frac{\partial x_1}{\partial y_1} & \frac{\partial x_1}{\partial y_2} \\ \frac{\partial x_2}{\partial y_1} & \frac{\partial x_2}{\partial y_2} \end{vmatrix}$$

the joint pdf of Y_1, Y_2 is,

$$g(y_1, y_2) = f(\underbrace{w_1(y_1, y_2)}_{X_1}, \underbrace{w_2(y_1, y_2)}_{X_2}) |J|$$

Let X_1 and X_2 have joint pdf

$$f(x_1, x_2) = 2e^{-(x_1+x_2)}, \quad 0 < x_1 < x_2$$



a) $Y_1 = X_2 - X_1, \quad Y_2 = X_1$

i) find w_1 and w_2

ii) Find J

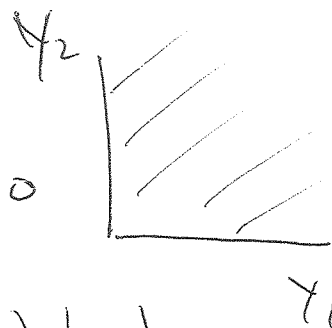
iii) Identify the support for Y_1 and Y_2

c) $X_1 = Y_2$
 $X_2 = Y_1 + Y_2$

ii) $J = \begin{vmatrix} 0 & 1 \\ 1 & 1 \end{vmatrix} = -1$

iii) • $X_1 > 0 \Rightarrow Y_2 > 0$

• $X_2 > X_1 \Rightarrow Y_1 + Y_2 > Y_2 \Rightarrow Y_1 > 0$



Finally, $g(y_1, y_2) = f(w_1(y_1, y_2), w_2(y_1, y_2)) |J|$
 $= 2e^{-(Y_1+2Y_2)} |-1|, \quad Y_1 > 0, Y_2 > 0$

Are Y_1 and Y_2 independent?

$g_1(y_1) = e^{-y_1}, y_1 > 0$ and $g_2(y_2) = 2e^{-2y_2}, y_2 > 0$

Y_1, Y_2 are independent

b) Let $z_1 = x_1 + x_2$, $z_2 = x_2/x_1$

i) Find inverse

$$x_1 = \frac{z_1}{1+z_2}$$

$$x_2 = \frac{z_1 z_2}{1+z_2}$$

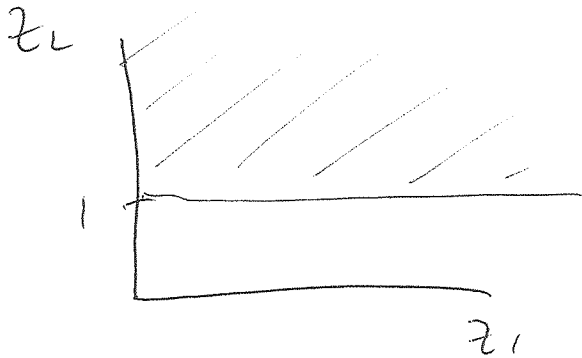
ii) ~~J~~ $J = \begin{vmatrix} \frac{1}{1+z_2} & -\frac{z_1}{(1+z_2)^2} \\ \frac{z_2}{1+z_2} & \frac{z_1(1+z_2) - z_1 z_2}{(1+z_2)^2} \end{vmatrix}$

$$J = \frac{z_1}{(1+z_2)^2}$$

iii) $0 < x_1 < x_2$

$$x_1 > 0 \Rightarrow \frac{z_1}{1+z_2} > 0 \Rightarrow \underline{z_1 > 0}$$

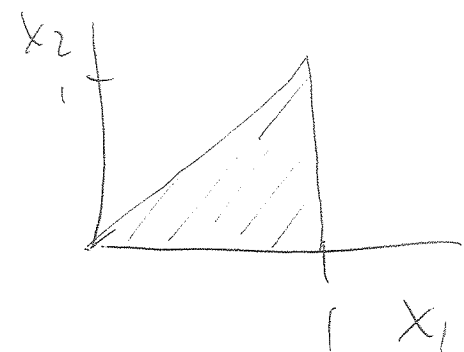
$$x_2 > x_1 \Rightarrow \frac{z_1 z_2}{1+z_2} > \frac{z_1}{1+z_2} \Rightarrow \underline{z_2 > 1}$$



$$g(z_1, z_2) = 2 e^{-z_1} \frac{z_1}{(1+z_2)^2}, \quad z_1 > 0, \quad z_2 > 1$$

2) x_1 and x_2 w/ joint pdf

$$f(x_1, x_2) = 15 x_1 x_2^2, \quad 0 < x_2 < x_1 < 1$$



a) $y_1 = x_1 + x_2, \quad y_2 = x_1$

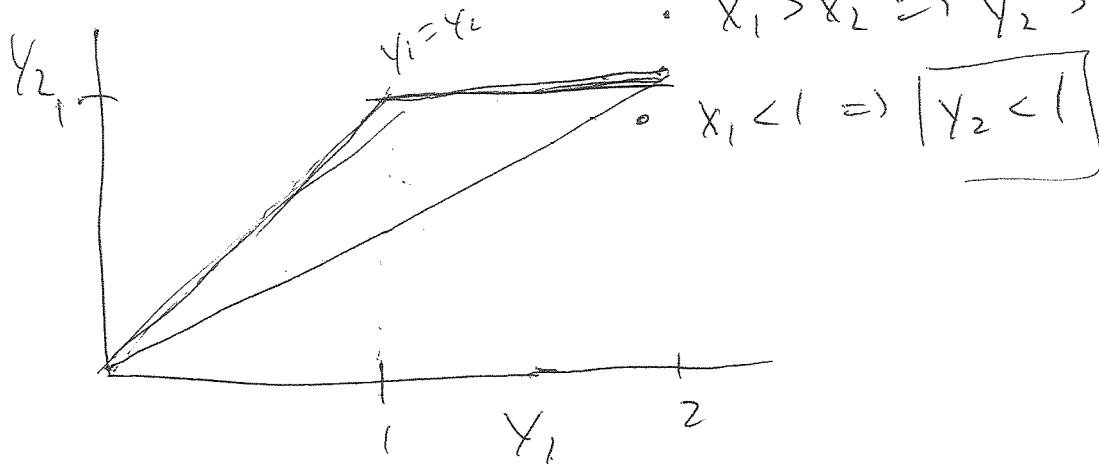
a i) $x_1 = y_2$

$$x_2 = y_1 - y_2$$

ii) $J = \begin{vmatrix} 0 & 1 \\ 1 & -1 \end{vmatrix} = -1$

iii) $0 < x_2 < x_1 < 1$ • $x_2 > 0 \Rightarrow y_1 - y_2 > 0 \Rightarrow \boxed{y_1 > y_2}$

• $x_1 > x_2 \Rightarrow y_2 > y_1 - y_2 \Rightarrow \boxed{y_2 > \frac{1}{2} y_1}$



$g(y_1, y_2) = 15 y_2 (y_1 - y_2)^2 |-1|, \quad \cancel{y_1 > y_2} \rightarrow y_1 > y_2 > \frac{1}{2} y_1, \quad y_2 < 1$

$$g(y_1) = \begin{cases} \int_{y_2=y_1}^{y_2=y_1} g(y_1, y_2) dy_2 & 0 < y_1 < 1 \\ \int_{y_2=y_1/2}^{y_2=y_1} g(y_1, y_2) dy_2 & 1 < y_1 < 2 \end{cases}$$