

uni: $g: \mathbb{R} \rightarrow \mathbb{R}$ what is dist of $g(x)$?

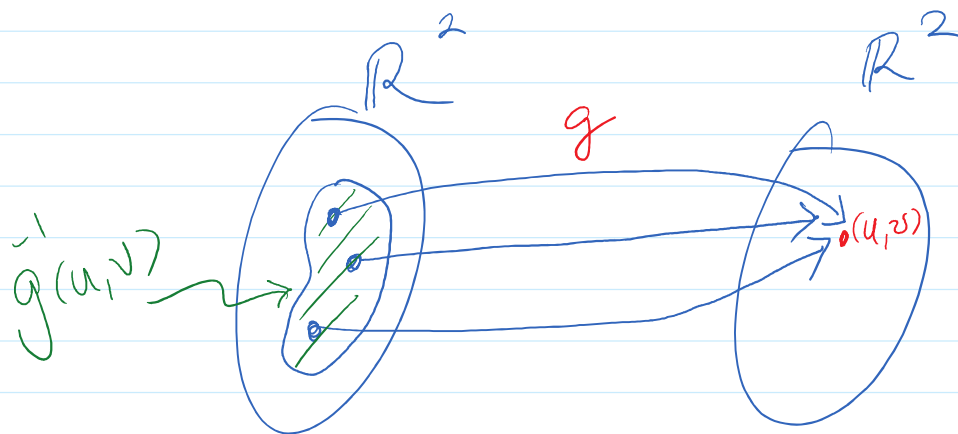
Biv: $g: \mathbb{R}^2 \rightarrow \mathbb{R}^2$ what is the dist of $g(x, y)$?

notation: $(x, y) \xrightarrow{g} (u, v)$

ex. $(u, v) = (\underbrace{x^2 y}_{g_1}, \underbrace{-\log y}_{g_2})$

Discrete: $(u, v) = (g_1(x, y), g_2(x, y))$

Assume x and y discrete.



$$g^{-1}(u, v) = \{(x, y) \mid g_1(x, y) = u \text{ and } g_2(x, y) = v\}$$

Want: PMF of (U, V) from PMF of (X, Y)

$$f_{u,v}(u,v) = P(U=u, V=v)$$

$$= P((U, V) \in \{(u, v)\})$$

$$= P(g(X, Y) \in \{(u, v)\})$$

$$= P((X, Y) \in g^{-1}(u, v))$$

same set

$$= \sum_{(x,y) \in g^{-1}(u,v)} f_{X,Y}(x,y)$$

Ex. let $X \perp Y$

and $X \sim \text{Pois}(\theta)$
 $Y \sim \text{Pois}(\lambda)$ } discrete

$$f_{X,Y}(x,y) = f_X(x) f_Y(y) = \frac{\theta^x e^{-\theta}}{x!} \frac{\lambda^y e^{-\lambda}}{y!}$$

let $U = X + Y$ and $V = Y$.

$$u = g_1(x,y) = x+y \quad \text{and} \quad v = g_2(x,y) = y$$

Solve for x, y in terms of u, v

$$\begin{cases} y = v = g_2^{-1}(u, v) \\ x = u - v = g_1^{-1}(u, v) \end{cases}$$

$$\begin{cases} u = x + y \\ x = u - y = u - v \end{cases}$$

So

$$f_{u,v}(u, v) = f_{x,y}(\underline{u-v}, \underline{v})$$

$$= \frac{\theta^{u-v} e^{-\theta}}{(u-v)!} \frac{\lambda^v e^{-\lambda}}{v!}$$

What is dist of u ?

$$u = x + y$$

$$v = y \quad \text{so } u \geq v$$

$$f_u(u) = \sum_{v=0}^u f(u, v)$$

$$= \frac{e^{-(\theta+\lambda)}}{u!} \sum_{v=0}^u \theta^{u-v} \lambda^v \frac{u!}{(u-v)! v!} \binom{u}{v}$$

$$= \frac{e^{-(\theta+\lambda)}}{u!} \sum_{v=0}^u \binom{u}{v} \lambda^v \theta^{u-v}$$

Binomial Theorem
 $(\theta + \lambda)^u$

$$f(u) = \frac{e^{-(\theta+\lambda)} (\theta+\lambda)^u}{u!}$$

$\curvearrowright \text{Pois}(\theta+\lambda)$

Theorem: $X \perp Y$ and $X \sim \text{Pois}(\theta)$
 $Y \sim \text{Pois}(\lambda)$

then $X + Y \sim \text{Pois}(\theta + \lambda)$.

What about cts?

If g is nice in uni case:

$$f_Y(y) = f_X(\underline{g^{-1}(y)}) \left| \frac{dg^{-1}}{dy} \right|$$

Bivariate Case:

- X and Y continuous
- $(u, v) = (g_1(x, y), g_2(x, y))$
- g is invertible and g^{-1} is differentiable

$$f_{u,v}(u,v) = f_{x,y}(g_1^{-1}(u,v), g_2^{-1}(u,v)) |\det J|$$

Jacobian of g^{-1}

Jacobians $h: \mathbb{R}^2 \rightarrow \mathbb{R}^2$

$$h(x,y) = (h_1(x,y), h_2(x,y))$$

then the Jacobian is

$$J = \begin{bmatrix} \frac{\partial h_1}{\partial x} & \frac{\partial h_1}{\partial y} \\ \frac{\partial h_2}{\partial x} & \frac{\partial h_2}{\partial y} \end{bmatrix}$$

in our case we want Jacobian of g^{-1}

$$J = \begin{bmatrix} \frac{\partial g_1^{-1}}{\partial u} & \frac{\partial g_1^{-1}}{\partial v} \\ \frac{\partial g_2^{-1}}{\partial u} & \frac{\partial g_2^{-1}}{\partial v} \end{bmatrix}$$

For a 2×2 Mtx

$$A = \begin{bmatrix} a & b \\ c & d \end{bmatrix} \text{ then } \det A = ad - cb$$

Process:

- ① get g_1^{-1}, g_2^{-1}
- ② Find J and $\det J$
- ③ plug in formula

Ex.
$$\begin{cases} (u, v) = (x+y, x-y) \\ u = g_1(x, y) = x+y \\ v = g_2(x, y) = x-y \end{cases}$$

Notice $u+v = 2x$

$$\text{So } x = \frac{u+v}{2}$$

and $u-v = 2y$

$$\text{So } y = \frac{u-v}{2}$$

① get inverses

$$x = g_1^{-1}(u, v) = \frac{u+v}{2}$$

$$y = g_2^{-1}(u, v) = \frac{u-v}{2}$$

② get J and $\det J$

$$J = \begin{bmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & -\frac{1}{2} \end{bmatrix}$$

$$\begin{aligned} \text{so } |\det J| &= \left| \left(\frac{1}{2}\right)\left(-\frac{1}{2}\right) - \left(\frac{1}{2}\right)\left(\frac{1}{2}\right) \right| \\ &= \left| -\frac{1}{2} \right| = \frac{1}{2} \end{aligned}$$

③ Plug in formula

$$f_{u,v}(u,v) = \underbrace{f_{X,Y}}_{\text{independent and identically distributed}}\left(\frac{u+v}{2}, \frac{u-v}{2}\right) \underline{\underline{\frac{1}{2}}}$$

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

independent and identically distributed

$$f_{X,Y}(x,y) = f_X(x) f_Y(y) = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}x^2} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}y^2}$$

$$f_{u,v}(u,v) = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{u+v}{2}\right)^2} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{u-v}{2}\right)^2} \frac{1}{2}$$

$$\begin{aligned} & \left[-\frac{1}{2} \left[\left(\frac{u+v}{2}\right)^2 + \left(\frac{u-v}{2}\right)^2 \right] \right] \\ &= -\frac{1}{2} \left(\frac{1}{4} (u^2 + \cancel{2uv} + v^2) + \frac{1}{4} (u^2 - \cancel{2uv} + v^2) \right) \\ &= -\frac{1}{2} \left(\frac{1}{4} \right) (2u^2 + 2v^2) \\ &= -\frac{1}{2} \left(\frac{1}{2} u^2 + \frac{1}{2} v^2 \right) \end{aligned}$$

$$= \frac{1}{2} \frac{1}{\sqrt{2\pi}} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(\frac{1}{2}u^2 + \frac{1}{2}v^2)}$$

$$f(u, v) = \underbrace{\frac{1}{\sqrt{2}} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(\frac{u^2}{2})}}_{N(0, 2)} \underbrace{\frac{1}{\sqrt{2}} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(\frac{v^2}{2})}}_{N(0, 2)}$$

only involves u

only v

$$u, v \stackrel{iid}{\sim} N(0, 2)$$

Ex.

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

$$Y \sim \text{Gamma}(\beta, \lambda)$$

and $X \perp Y$

$$U = \frac{X + Y}{g_1} \quad \text{and} \quad V = \frac{\cancel{X}}{\cancel{X} + Y} \quad g_2$$

① get inverses

$$uv = (x+y) \left(\frac{x}{x+y} \right) = x$$

$$\text{so } \boxed{x = g_1^{-1}(u, v) = uv}$$

$$u = x + y \quad \text{so} \quad y = u - x = u - uv = u(1-v)$$

$$\text{so } \boxed{f(u, v) = \frac{1}{g_1 g_2} f(x, y)}$$

so $\boxed{y = g_2^{-1}(u, v) = u(1-v)}$

② Get J and $\det J$

$$J = \begin{bmatrix} v & u \\ 1-v & -u \end{bmatrix} \Rightarrow \det J = -uv - u(1-v) = -u$$

$$\Rightarrow |\det J| = u$$

③ plug-in

$$f_{X,Y}(x,y) = f_X(x) f_Y(y) = \frac{\lambda e^{-\lambda x} (\lambda x)^{\alpha-1}}{\Gamma(\alpha)} \frac{\lambda e^{-\lambda y} (\lambda y)^{\beta-1}}{\Gamma(\beta)}$$

$$f(u,v) = f_{X,Y}(uv, u(1-v)) u$$

$$= \frac{\lambda e^{-\lambda uv} (\lambda uv)^{\alpha-1}}{\Gamma(\alpha)} \frac{\lambda e^{-\lambda u(1-v)} (u(1-v)\lambda)^{\beta-1}}{\Gamma(\beta)} u$$

= ... algebra

$$f(u,v) = \frac{\lambda^{\alpha+\beta}}{\Gamma(\alpha)\Gamma(\beta)} \underbrace{u^{\alpha+\beta-1} e^{-\lambda u}}_{\text{only } u \text{ (some Gamma)}} \underbrace{v^{\alpha-1} (1-v)^{\beta-1}}_{\text{only } v \text{ (some Beta)}}$$

So $U \perp V$.

and $U \sim \text{Gamma}(\alpha+\beta, \lambda)$

$V \sim \text{Beta}(\alpha, \beta)$

Theorem: Independence and Transformations

If $X \perp Y$ and

$$U = g(X) \leftarrow \text{only } X$$

$$V = h(Y) \leftarrow \text{only } Y$$

then $U \perp V$.

Ex. $U = X^2$ and $V = \log Y$.

Ex. $U = XY$ and $V = X$

what is $\text{dist}(u,v)$?

① get inverses

$$u = xy \quad \text{and} \quad v = x$$

$$x = g_1^{-1}(u, v) = v$$

$$y = g_2^{-1}(u, v) = u/x = u/v$$

② get J and $\det J$

$$J = \begin{bmatrix} 0 & 1 \\ 1/v & -u/v^2 \end{bmatrix}$$

$$\begin{aligned} \text{then } \det J &= (0)(-u/v^2) - (1/v)(1) \\ &= -1/v \end{aligned}$$

$$\text{hence } |\det J| = 1/v \quad (\text{assuming } v > 0)$$

③ plug-in

$$f(u, v) = f_{x, y}(v, u/v) \frac{1}{v}$$