

**SOR201****Solutions to Examples 6**

1. (i) The c.d.f. of the r.v.  $X$  is  $F(x) = P(X \leq x)$ ,  $-\infty < x < \infty$ .

(a) The event  $(-\infty < X \leq x) \subseteq$  the event  $(-\infty < X \leq y)$  when  $x \leq y$ .

$$\text{So } P(-\infty < X \leq x) \leq P(-\infty < X \leq y), \quad x \leq y$$

i.e.  $F(x) \leq F(y)$  when  $x \leq y$ .

$$(b) \quad F(-\infty) = P(X \leq -\infty) = 0,$$

$$F(+\infty) = P(X \leq +\infty) = 1.$$

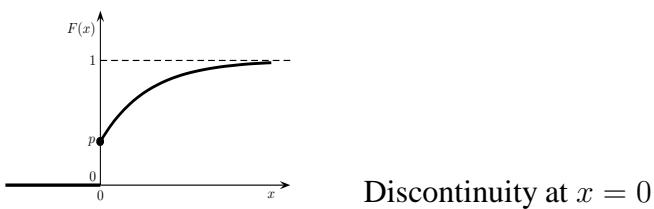
$$(c) \quad (-\infty < X \leq b) = (-\infty < X \leq a) \cup (a < X \leq b)$$

– the union of 2 m.e. events.

$$\text{So } P(-\infty < X \leq b) = P(-\infty < X \leq a) + P(a < X \leq b)$$

$$\begin{aligned} \text{i.e. } P(a < X \leq b) &= P(-\infty < X \leq b) - P(-\infty < X \leq a) \\ &= F(b) - F(a). \end{aligned}$$

(ii) (a)



Discontinuity at  $x = 0$ .

(b) The c.d.f. is

$$F(x) = \begin{cases} 0, & x < 0 \\ p, & x = 0 \\ p + k(1 - e^{-\lambda x}) = p + (1-p)(1 - e^{-\lambda x}), & x > 0. \end{cases}$$

[ $k$  determined as  $(1-p)$  by requiring that  $F(x) \rightarrow 1$  as  $x \rightarrow \infty$ ].

(iii)  $f(x)$  is symmetrical about the point  $a$ , i.e.

$$f(a-y) = f(a+y), \quad y \geq 0. \quad (**)$$

Then

$$\begin{aligned} \int_{-\infty}^a f(x) dx &= \int_0^0 f(a+y) dy \quad [\text{setting } x = a-y] \\ &= \int_0^\infty f(a-y) dy \end{aligned}$$

and

$$\begin{aligned} \int_a^\infty f(x) dx &= \int_{-\infty}^a f(a-y)(-1) dy \quad [\text{setting } x = a+y] \\ &= \int_0^\infty f(a-y) dy \quad [\text{using } (**)] \end{aligned}$$

$$\text{So } \int_{-\infty}^a f(x) dx = \int_a^\infty f(x) dx.$$

$$\text{But } \int_{-\infty}^a f(x) dx + \int_a^\infty f(x) dx = \int_{-\infty}^\infty f(x) dx = 1.$$

So

$$\int_{-\infty}^a f(x) dx = \int_a^\infty f(x) dx = \frac{1}{2}$$

$$\text{i.e. } P(X \leq a) = P(X \geq a) = \frac{1}{2}$$

i.e. the median of  $X$  is  $a$ .

/continued overleaf

Also

$$\begin{aligned}
 E(X) &= \int_{-\infty}^{\infty} xf(x)dx \\
 &= \int_{-\infty}^{-a} xf(x)dx + \int_a^{\infty} xf(x)dx \\
 &\quad [\text{set } y = a - x \text{ in first integral, } y = x - a \text{ in second}] \\
 &= \int_0^0 (a - y)f(a - y)(-1)dy + \int_0^{\infty} (a + y)f(a + y)dy \\
 &= \int_0^{\infty} (a - y)f(a - y)dy + \int_0^{\infty} (a + y)f(a - y)dy \quad [\text{using } (**)] \\
 &= \int_0^{\infty} 2af(a - y)dy \\
 &\quad [\text{use } \int_0^{\infty} f(a - y)dy = \int_a^{\infty} f(x)dx = \frac{1}{2}] \\
 &= 2a \cdot \frac{1}{2} = a.
 \end{aligned}$$

2. (i) The c.d.f. is

$$F(x) = \begin{cases} 1 - \exp(-\frac{1}{2}x^2), & x \geq 0 \\ 0, & x < 0. \end{cases}$$

So

$$f(x) = \frac{dF(x)}{dx} = \begin{cases} x \exp(-\frac{1}{2}x^2), & x \geq 0 \\ 0, & x < 0. \end{cases}$$

Then

$$\begin{aligned}
 E(X) &= \int_{-\infty}^{\infty} xf(x)dx \\
 &= \int_0^{\infty} x^2 e^{-\frac{1}{2}x^2} dx \quad [\text{set } t = \frac{1}{2}x^2, dt = xdx = \sqrt{2t}dx] \\
 &= \int_0^{\infty} 2te^{-t} \frac{1}{\sqrt{2t}} dt \\
 &= \sqrt{2} \int_0^{\infty} t^{\frac{3}{2}-1} e^{-t} dt \\
 &= \sqrt{2}\Gamma(\frac{3}{2}) = \sqrt{2} \cdot \frac{1}{2} \cdot \Gamma(\frac{1}{2}) = \frac{1}{\sqrt{2}} \cdot \sqrt{\pi} = \sqrt{\pi/2}.
 \end{aligned}$$

$$\begin{aligned}
 E(X^2) &= \int_0^{\infty} x^3 e^{-\frac{1}{2}x^2} dx \quad [\text{set } t = \frac{1}{2}x^2] \\
 &= \int_0^{\infty} (2t)^{\frac{3}{2}} e^{-t} \frac{1}{\sqrt{2t}} dt \\
 &= 2 \int_0^{\infty} te^{-t} dt = 2\Gamma(2) = 2.
 \end{aligned}$$

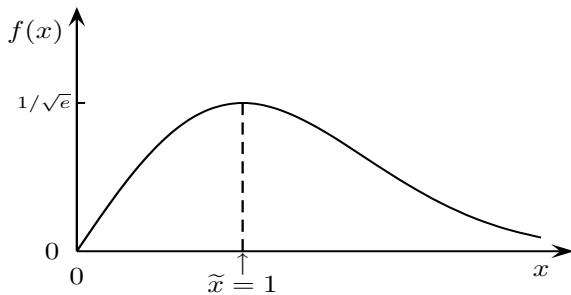
$$\text{So } \text{Var}X = E(X^2) - [E(X)]^2 = 2 - \pi/2.$$

The median  $m$  of a continuous distribution is such that  $F(m) = \frac{1}{2}$ , so here

$$\begin{aligned}
 \text{i.e.} \quad 1 - \exp(-\frac{1}{2}m^2) &= \frac{1}{2} \\
 \text{so} \quad \exp(-\frac{1}{2}m^2) &= \frac{1}{2} \quad \text{or} \quad \exp(\frac{1}{2}m^2) = 2, \\
 \text{i.e.} \quad \frac{1}{2}m^2 &= \log_e 2 \\
 &\quad m = \{2 \log_e 2\}^{\frac{1}{2}}.
 \end{aligned}$$

/continued overleaf

Also  $\frac{df(x)}{dx} = e^{-\frac{1}{2}x^2} - x^2e^{-\frac{1}{2}x^2} = 0$  when  $x = \pm 1$ . So mode  $\tilde{x} = 1$ .



(ii)

$$f_X(x) = \begin{cases} kx^{p-1}/(1+x)^{p+q}, & x \geq 0; p, q > 0 \\ 0, & \text{otherwise.} \end{cases}$$

The transformation  $y = 1/(1+x)$ ,  $x \geq 0$  is one-to-one and differentiable:

$$\frac{dy}{dx} = -\frac{1}{(1+x)^2} = -y^2.$$

The inverse transformation is

$$x = \frac{(1-y)}{y}, \quad \frac{1-y}{y} \geq 0 \quad \text{or} \quad 0 < y \leq 1.$$

So

$$\begin{aligned} f_Y(y) &= f_X\left(\frac{1-y}{y}\right) \left| \frac{dx}{dy} \right| \\ &= \begin{cases} k\left(\frac{1-y}{y}\right)^{p-1} y^{p+q} \left| -\frac{1}{y^2} \right|, & 0 < y \leq 1 \\ 0, & \text{otherwise} \end{cases} \\ \text{i.e. } f_Y(y) &= \begin{cases} ky^{q-1}(1-y)^{p-1}, & 0 < y \leq 1 \\ 0, & \text{otherwise} \end{cases} \\ \text{i.e. } Y &\sim \text{beta}(q, p). \end{aligned}$$

(iii)

$$f_X(x) = \begin{cases} 1, & 0 \leq x \leq 1 \\ 0, & \text{otherwise.} \end{cases}$$

The transformation  $y = \log_e \{1/(1-x)\}$ ,  $0 \leq x \leq 1$  is one-to-one and differentiable. The inverse transformation is

$$\begin{aligned} x &= 1 - e^{-y}, \quad 0 \leq 1 - e^{-y} \leq 1, \quad \text{i.e. } 0 \leq e^{-y} \leq 1, \quad \text{i.e. } 0 \leq y < \infty \\ \text{and } \frac{dx}{dy} &= e^{-y}. \end{aligned}$$

So

$$\begin{aligned} f_Y(y) &= f_X(1 - e^{-y}) \left| \frac{dx}{dy} \right| \\ &= \begin{cases} e^{-y}, & 0 \leq y < \infty \\ 0, & \text{otherwise} \end{cases} \end{aligned}$$

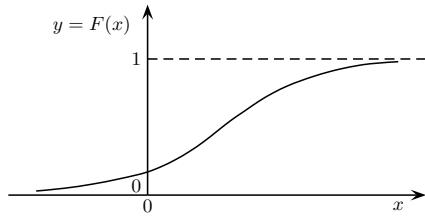
– the negative exponential distribution with  $\lambda = 1$ .

/continued overleaf

- (iv) The transformation  $y = F_X(x)$ ,  $-\infty < x < \infty$  is one-to-one and differentiable, and

$$\frac{dy}{dx} = f_X(x), \quad 0 \leq y \leq 1.$$

So



$$f_Y(y) = f_X\{F_X^{-1}(y)\} \left| \frac{dx}{dy} \right| = f_X(F_X^{-1}(y)) \cdot \frac{1}{f_X(F_X^{-1}(y))} = 1, \quad 0 \leq y \leq 1,$$

i.e.  $Y \sim \text{uniform}[0, 1]$ .

3. (i) (a)  $f_X(x) = 2x \exp(-x^2)$ ,  $0 \leq x \leq \infty$ .

The transformation  $y = x^2$ ,  $0 \leq x < \infty$  is one-to-one and differentiable.

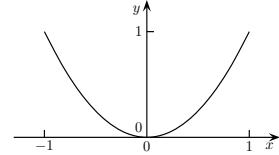
Its inverse is  $x = +\sqrt{y}$  and  $\frac{dy}{dx} = 2x = 2\sqrt{y}$ ,  $0 \leq y < \infty$ .

So

$$\begin{aligned} f_Y(y) &= f_X(\sqrt{y}) \left| \frac{dx}{dy} \right| \\ &= 2\sqrt{y} e^{-y} \left| \frac{1}{2\sqrt{y}} \right| = e^{-y}, \quad 0 \leq y < \infty \end{aligned}$$

(b)

$$f_X(x) = \begin{cases} \frac{1}{2}(1+x), & -1 \leq x \leq 1 \\ 0, & \text{otherwise.} \end{cases}$$



The transformation  $y = x^2$  is not one-to-one when  $-1 \leq x \leq 1$ . Then  
**EITHER**

$$\begin{aligned} F_Y(y) &= P(Y \leq y) = P(X^2 \leq y) = P(-\sqrt{y} \leq X \leq +\sqrt{y}) \\ &= \int_{-\sqrt{y}}^{\sqrt{y}} f_X(x) dx = \int_{-\sqrt{y}}^{\sqrt{y}} \frac{1}{2}(1+x) dx, \quad -1 \leq -\sqrt{y} \leq \sqrt{y} \leq 1 \\ &= \left[ \frac{1}{2}x + \frac{1}{4}x^2 \right]_{-\sqrt{y}}^{\sqrt{y}} = \sqrt{y}, \quad 0 \leq y \leq 1 \end{aligned}$$

So

$$f_Y(y) = \begin{cases} \frac{1}{2\sqrt{y}}, & 0 \leq y \leq 1 \\ 0, & \text{otherwise.} \end{cases}$$

**OR**

Divide the interval  $(-1, 1)$  into sub-intervals such that the transformation  $y = x^2$  is one-to-one in each sub-interval. Thus

$$\underline{-1 \leq x \leq 0 :} \quad y = x^2 \Rightarrow x = -\sqrt{y}, \quad \frac{dy}{dx} = 2x = -2\sqrt{y};$$

$$f_Y^-(y) = f_X^-(-\sqrt{y}) \left| \frac{dx}{dy} \right| = \frac{1}{2}(1-\sqrt{y}) \cdot \frac{1}{2\sqrt{y}}, \quad -1 \leq -\sqrt{y} \leq 0, \quad \text{i.e. } 0 \leq \sqrt{y} \leq 1.$$

$$\underline{0 \leq x \leq 1 :} \quad y = x^2 \Rightarrow x = +\sqrt{y}, \quad \frac{dy}{dx} = 2x = 2\sqrt{y};$$

$$f_Y^+(y) = f_X^+(+\sqrt{y}) \left| \frac{dx}{dy} \right| = \frac{1}{2}(1+\sqrt{y}) \cdot \frac{1}{2\sqrt{y}}, \quad 0 \leq \sqrt{y} \leq 1.$$

/continued overleaf

Then

$$\begin{aligned} f_Y(y) &= f_Y^+(y) + f_Y^-(y) \\ &= \begin{cases} \frac{1}{2\sqrt{y}}, & 0 \leq y \leq 1 \\ 0, & \text{otherwise.} \end{cases} \end{aligned}$$

$$(c) \quad f_X(x) = \begin{cases} \frac{1}{2}, & -\frac{1}{2} \leq x \leq \frac{3}{2} \\ 0, & \text{otherwise.} \end{cases}$$

The transformation  $y = x^2$  is *not* one-to-one when  $-\frac{1}{2} \leq x \leq \frac{3}{2}$ . So  
**EITHER**

$$P(Y \leq y) = P(X^2 \leq y) = P(-\sqrt{y} \leq X \leq +\sqrt{y}) = \int_{-\sqrt{y}}^{+\sqrt{y}} f_X(x) dx.$$

Now

$$\begin{aligned} \int_{-\sqrt{y}}^{\sqrt{y}} f_X(x) dx &= \int_{-\sqrt{y}}^{\sqrt{y}} \frac{1}{2} dx \quad \underline{\text{provided}} \quad -\frac{1}{2} \leq -\sqrt{y} \leq \sqrt{y} \leq \frac{1}{2} \\ &\quad \text{i.e. } 0 \leq y \leq \frac{1}{4} \\ &= \left[ \frac{1}{2}x \right]_{-\sqrt{y}}^{\sqrt{y}} = \sqrt{y}, \quad 0 \leq y \leq \frac{1}{4} \end{aligned}$$

while

$$\begin{aligned} \int_{-\sqrt{y}}^{\sqrt{y}} f_X(x) dx &= \int_{-\frac{1}{2}}^{\frac{1}{2}} \frac{1}{2} dx \quad \underline{\text{provided}} \quad -\frac{3}{2} \leq -\sqrt{y} \leq -\frac{1}{2} \leq \frac{1}{2} < \sqrt{y} \leq \frac{3}{2} \\ &\quad \text{i.e. } \frac{1}{4} < y \leq \frac{9}{4} \\ &= \left[ \frac{1}{2}x \right]_{-\frac{1}{2}}^{\frac{1}{2}} = \frac{1}{2}\sqrt{y} + \frac{1}{4}, \quad \frac{1}{4} < y \leq \frac{9}{4} \end{aligned}$$

so that

$$F_Y(y) = \begin{cases} \sqrt{y}, & 0 \leq y \leq \frac{1}{4} \\ \frac{1}{2}\sqrt{y} + \frac{1}{4}, & \frac{1}{4} < y \leq \frac{9}{4}. \end{cases}$$

Hence

$$f_Y(y) = \begin{cases} \frac{1}{2\sqrt{y}}, & 0 \leq y \leq \frac{1}{4} \\ \frac{1}{4\sqrt{y}}, & \frac{1}{4} < y \leq \frac{9}{4}. \end{cases}$$

**OR**

$$\underline{-\frac{1}{2} \leq x \leq 0 :} \quad y = x^2 \Rightarrow x = -\sqrt{y}, \quad \frac{dy}{dx} = 2x = -2\sqrt{y};$$

$$f_Y^-(y) = f_X^-(-\sqrt{y}) \left| \frac{dx}{dy} \right| = \frac{1}{2} \cdot \frac{1}{2\sqrt{y}}, \quad -\frac{1}{2} \leq -\sqrt{y} \leq 0 \text{ or } 0 \leq y \leq \frac{1}{4}.$$

$$\underline{0 \leq x \leq \frac{3}{2} :} \quad y = x^2 \Rightarrow x = +\sqrt{y}, \quad \frac{dy}{dx} = 2x = 2\sqrt{y};$$

$$f_Y^+(y) = f_X^+(+\sqrt{y}) \left| \frac{dx}{dy} \right| = \frac{1}{2} \cdot \frac{1}{2\sqrt{y}}, \quad 0 \leq \sqrt{y} \leq \frac{3}{2} \text{ or } 0 \leq y \leq \frac{9}{4}.$$

Then

$$f_Y(y) = f_Y^+(y) + f_Y^-(y)$$

$$= \begin{cases} \frac{1}{2\sqrt{y}}, & 0 \leq y \leq \frac{1}{4} \\ \frac{1}{4\sqrt{y}}, & \frac{1}{4} < y \leq \frac{9}{4}. \end{cases}$$

/continued overleaf

$$(ii) \quad f_Z(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}z^2}, \quad -\infty < z < \infty.$$

We note that the transformation  $v = z^2$  is *not* one-to-one when  $-\infty < z < \infty$ ;  $0 \leq v < \infty$ . Then

**EITHER**

$$\begin{aligned} F_V(v) &= P(V \leq v) = P(Z^2 \leq v) \\ &= P(-\sqrt{v} \leq Z \leq +\sqrt{v}) = F_Z(+\sqrt{v}) - F_Z(-\sqrt{v}). \end{aligned}$$

So

$$\begin{aligned} f_V(v) &= \frac{dF_V(v)}{dv} = \frac{d}{dv} \{F_Z(\sqrt{v}) - F_Z(-\sqrt{v})\} \\ &= f_Z(\sqrt{v}) \cdot \frac{1}{2\sqrt{v}} - f_Z(-\sqrt{v}) \left( -\frac{1}{2\sqrt{v}} \right) \\ &= \frac{1}{2\sqrt{v}} \left\{ \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}v} + \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}v} \right\} \\ &= \frac{1}{2^{\frac{1}{2}} \sqrt{\pi}} v^{-\frac{1}{2}} e^{-\frac{1}{2}v} \\ &= \frac{1}{2^{\frac{1}{2}} \Gamma(\frac{1}{2})} v^{\frac{1}{2}-1} e^{-\frac{1}{2}v}, \quad 0 \leq v < \infty \end{aligned}$$

which is the p.d.f. of the  $\chi^2(1)$  distribution,

i.e.  $V \sim \chi^2(1)$ .

**OR**

Consider the one-to-one transformations for  $-\infty < z < 0$  and  $0 \leq z < \infty$ .

$$\underline{-\infty < z < 0}: \quad v = z^2 \Rightarrow z = -\sqrt{v}, \quad \frac{dv}{dz} = 2z = -2\sqrt{v};$$

$$\therefore f_V^-(v) = f_Z^-(-\sqrt{v}) \left| \frac{dz}{dv} \right| = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}v} \frac{1}{2\sqrt{v}}, \quad 0 \leq v < \infty.$$

$$\underline{0 \leq z < \infty}: \quad v = z^2 \Rightarrow z = +\sqrt{v}, \quad \frac{dv}{dz} = 2z = 2\sqrt{v};$$

$$\therefore f_V^+(v) = f_Z^+(+\sqrt{v}) \left| \frac{dz}{dv} \right| = \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}v} \frac{1}{2\sqrt{v}}, \quad 0 \leq v < \infty.$$

$$\text{So } f_V(v) = f_V^+(v) + f_V^-(v) = \frac{2}{\sqrt{2\pi}} e^{-\frac{1}{2}v} \frac{1}{2\sqrt{v}} \longrightarrow \text{ above result.}$$

$$(iii) \quad f_X(x) = \begin{cases} 1, & 0 \leq x \leq 1 \\ 0, & \text{otherwise.} \end{cases}$$

The transformation  $y = x^\alpha$  is one-to-one when  $0 \leq x \leq 1$ , and

$$\frac{dy}{dx} = \alpha x^{\alpha-1} = \alpha y^{(\alpha-1)/\alpha}.$$

So

$$\begin{aligned} f_Y(y) &= f_X(y^{1/\alpha}) \left| \frac{dx}{dy} \right| = \begin{cases} 1 \cdot \left| \frac{1}{\alpha y^{(1-\alpha)/\alpha}} \right|, & 0 \leq y^{1/\alpha} \leq 1 \\ 0, & \text{otherwise} \end{cases} \\ &= \begin{cases} \frac{1}{\alpha} y^{(1/\alpha)-1}, & 0 \leq y \leq 1; \alpha > 0 \\ -\frac{1}{\alpha} y^{(1/\alpha)-1}, & 1 \leq y < \infty; \alpha < 0. \end{cases} \end{aligned}$$

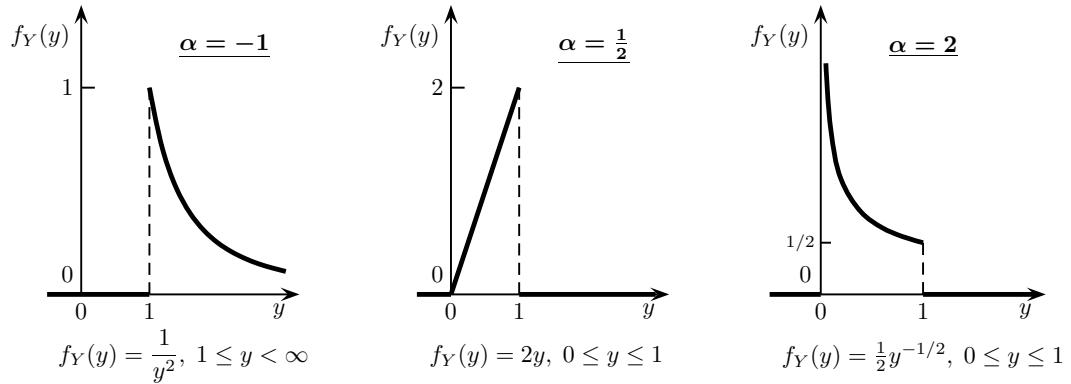
For  $\alpha > 0$ :

$$F_Y(y) = \begin{cases} 0, & y < 0 \\ 1, & y > 1 \\ \int_0^y \frac{1}{\alpha} t^{(1/\alpha)-1} dt = y^{1/\alpha}, & 0 \leq y \leq 1. \end{cases}$$

For  $\alpha < 0$ :

$$F_Y(y) = \begin{cases} 0, & y < 1 \\ \int_1^y (-\frac{1}{\alpha}) t^{(1/\alpha)-1} dt = y^{1/\alpha}, & y \geq 1. \end{cases}$$

For the three cases asked for, we then have:



#### 4. (i) Normal distribution $N(\mu, \sigma^2)$

(a)

$$\begin{aligned} \int_{-\infty}^{\infty} f_X(x) dx &= \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}\sigma} \exp\left\{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2\right\} dx \quad [\text{set } z = \frac{x-\mu}{\sigma}] \\ &= \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}z^2} dz \\ &= \int_{-\infty}^0 \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}z^2} dz + \int_0^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}z^2} dz \\ &= \int_{-\infty}^0 \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}y^2} (-dy) + \quad \text{ditto} \quad [\text{setting } z = -y] \\ &= 2 \int_0^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}y^2} dy \quad [\text{set } t = \frac{1}{2}y^2, y = \sqrt{2t}, \\ &\quad dt = ydy = \sqrt{2t}dy] \\ &= \frac{2}{\sqrt{2\pi}} \int_0^{\infty} e^{-t} \frac{1}{\sqrt{2t}} dt \\ &= \frac{1}{\sqrt{\pi}} \int_0^{\infty} t^{\frac{1}{2}-1} e^{-t} dt = \frac{1}{\sqrt{\pi}} \Gamma(\frac{1}{2}) = 1. \quad [\text{using } \Gamma(\frac{1}{2}) = \sqrt{\pi}] \end{aligned}$$

/continued overleaf

(b) The transformation  $w = a + bx$  is one-to-one and differentiable;

$$\frac{dw}{dx} = b, \quad x = \frac{w-a}{b}.$$

Then

$$\begin{aligned} f_W(w) &= f_X\left(\frac{w-a}{b}\right) \left| \frac{dx}{dw} \right| \\ &= \frac{1}{\sqrt{2\pi}\sigma} \exp\left\{-\frac{1}{2}\left(\frac{\frac{w-a}{b}-\mu}{\sigma}\right)^2\right\} \left| \frac{1}{b} \right|, \quad -\infty < \frac{w-a}{b} < \infty \\ &= \frac{1}{\sqrt{2\pi}\sigma|b|} \exp\left\{-\frac{1}{2}\left(\frac{w-a-\mu b}{b\sigma}\right)^2\right\}, \quad -\infty < w < \infty \end{aligned}$$

i.e.  $W \sim N(a + b\mu, b^2\sigma^2)$ , and  $SD(W) = \sqrt{b^2\sigma^2} = |b|\sigma$ .

(ii) Negative exponential distribution ( $\lambda$ )

(a)

$$\begin{aligned} E(X) &= \int_0^\infty x \cdot \lambda e^{-\lambda x} dx \quad [\text{set } t = \lambda x, dt = \lambda dx] \\ &= \int_0^\infty \left(\frac{t}{\lambda}\right) \lambda e^{-t} \cdot \frac{1}{\lambda} dt = \frac{1}{\lambda} \int_0^\infty t^{2-1} e^{-t} dt \\ &= \frac{1}{\lambda} \Gamma(2) = \frac{1}{\lambda} 1! = \frac{1}{\lambda}. \end{aligned}$$

Similarly  $E(X^2) = \frac{1}{\lambda^2} \Gamma(3) = \frac{1}{\lambda^2} 2! = \frac{2}{\lambda^2}$ .

So  $\text{Var}(X) = \frac{2}{\lambda^2} - \frac{1}{\lambda^2} = \frac{1}{\lambda^2}$ .

(b)

$$\begin{aligned} P(X > s+t | X > s) &= \frac{P(X > s+t)}{P(X > s)} \quad \text{when } t > 0 \\ &= \frac{e^{-\lambda(s+t)}}{e^{-\lambda s}} \quad [\text{since } P(X > y) = 1 - P(X \leq y) \\ &\quad = 1 - \{1 - e^{-\lambda y}\} = e^{-\lambda y}] \\ &= e^{-\lambda t} = P(X > t). \end{aligned}$$

(iii) Gamma distribution ( $\alpha, \lambda$ )

(a)

$$\begin{aligned} E(X) &= \int_0^\infty x \frac{\lambda^\alpha x^{\alpha-1} \exp(-\lambda x)}{\Gamma(\alpha)} dx \quad [\text{set } t = \lambda x, dt = \lambda dx] \\ &= \int_0^\infty \left(\frac{t}{\lambda}\right) \frac{t^{\alpha-1} e^{-t}}{\Gamma(\alpha)} dt \\ &= \frac{1}{\lambda \Gamma(\alpha)} \int_0^\infty t^{(\alpha+1)-1} e^{-t} dt \\ &= \frac{\Gamma(\alpha+1)}{\lambda \Gamma(\alpha)} = \frac{\alpha}{\lambda} \quad [\text{since } \Gamma(\alpha+1) = \alpha \Gamma(\alpha)]. \end{aligned}$$

Similarly  $E(X^2) = \frac{\Gamma(\alpha+2)}{\lambda^2 \Gamma(\alpha)} = \frac{(\alpha+1)\alpha}{\lambda^2}$ .

So  $\text{Var}(X) = \frac{(\alpha+1)\alpha}{\lambda^2} - \left(\frac{\alpha}{\lambda}\right)^2 = \frac{\alpha}{\lambda^2}$ .

/continued overleaf

(b)  $V \sim \chi^2(r) = \text{Gamma } (\alpha = \frac{r}{2}, \lambda = \frac{1}{2}).$

$$\text{So } f_V(v) = \frac{v^{\frac{r}{2}-1} \exp(-\frac{v}{2})}{2^{\frac{r}{2}} \Gamma(\frac{r}{2})}, \quad v \geq 0.$$

$$\mathbb{E}(V) = \frac{\frac{r}{2}}{\frac{1}{2}} = r; \quad \text{Var}(V) = \frac{\frac{r}{2}}{(\frac{1}{2})^2} = 2r.$$

(iv) Beta distribution  $(a, b)$

(a)

$$\begin{aligned} \mathbb{E}(X) &= \int_0^1 x \cdot \frac{1}{B(a, b)} x^{a-1} (1-x)^{b-1} dx \\ &= \frac{1}{B(a, b)} \int_0^1 x^a (1-x)^{b-1} dx \\ &= \frac{B(a+1, b)}{B(a, b)} = \frac{\Gamma(a+1)\Gamma(b)}{\Gamma(a+b+1)} \cdot \frac{\Gamma(a+b)}{\Gamma(a)\Gamma(b)} = \frac{a}{a+b}. \end{aligned}$$

Similarly

$$\begin{aligned} \mathbb{E}(X^2) &= \frac{B(a+2, b)}{B(a, b)} \\ &= \frac{\Gamma(a+2)\Gamma(b)}{\Gamma(a+b+2)} \cdot \frac{\Gamma(a+b)}{\Gamma(a)\Gamma(b)} = \frac{(a+1)a}{(a+b+1)(a+b)}. \end{aligned}$$

So

$$\begin{aligned} \text{Var}(X) &= \frac{(a+1)a}{(a+b+1)(a+b)} - \left( \frac{a}{a+b} \right)^2 \\ &= \frac{(a+1)a(a+b) - a^2(a+b+1)}{(a+b)^2(a+b+1)} \\ &= \frac{ab}{(a+b)^2(a+b+1)}. \end{aligned}$$

(b) The transformation  $y = 1 - x$  is one-to-one and differentiable;

$$x = 1 - y; \quad \frac{dy}{dx} = -1$$

Then

$$\begin{aligned} f_Y(y) &= f_X(1-y)|-1|, \quad 0 \leq y \leq 1 \\ &= \frac{1}{B(a, b)} (1-y)^{a-1} y^{b-1}, \quad 0 \leq y \leq 1 \end{aligned}$$

i.e.

$$Y \sim \text{beta } (b, a).$$

(c) Transforming  $0 \leq x \leq 1$  to  $A \leq w \leq B$  involves a *location* change  $0 \rightarrow A$  and a *scale* change of  $1 - 0 \rightarrow B - A$ , i.e. a linear transformation

$$w = A + (B - A)x.$$

This is a one-to-one differentiable transformation;

$$x = (w - A)/(B - A); \quad \frac{dw}{dx} = B - A.$$

So

$$\begin{aligned} f_W(w) &= f_X \left\{ \frac{w - A}{B - A} \right\} \left| \frac{dx}{dw} \right| \\ &= \frac{1}{B(a, b)} \left( \frac{w - A}{B - A} \right)^{a-1} \left( 1 - \frac{w - A}{B - A} \right)^{b-1} \cdot \frac{1}{B - A}, \quad A \leq w \leq B \\ &= \frac{1}{B(a, b)} \cdot \frac{(w - A)^{a-1} (B - w)^{b-1}}{(B - A)^{a+b-1}}, \quad A \leq w \leq B. \end{aligned}$$

(v) Weibull distribution  $(b, c)$ 

$$(a) F(x) = 1 - \exp \left\{ - \left( \frac{x}{b} \right)^c \right\}, \quad x \geq 0.$$

$$\text{So } f_X(x) = \frac{dF}{dx} = \frac{cx^{c-1}}{b^c} \exp \left\{ - \left( \frac{x}{b} \right)^c \right\}, \quad x \geq 0.$$

$$\begin{aligned} E(X) &= \int_0^\infty x \frac{cx^{c-1}}{b^c} \exp \left\{ - \left( \frac{x}{b} \right)^c \right\} dx \quad [\text{set } t = (\frac{x}{b})^c, dt = \frac{cx^{c-1}}{b^c} dx, x = bt^{1/c}] \\ &= \int_0^\infty bt^{1/c} e^{-t} dt = b\Gamma(1 + 1/c). \end{aligned}$$

$$(b) \text{ Survival function } \bar{F}(x) = 1 - F(x) = \exp \left\{ - \left( \frac{x}{b} \right)^c \right\}, \quad x \geq 0.$$

*Hazard function*

$$H(x) = -\log_e(1 - F(x)) = -\log_e \left[ \exp \left\{ - \left( \frac{x}{b} \right)^c \right\} \right] = \left( \frac{x}{b} \right)^c, \quad x \geq 0.$$

$$\text{Hazard rate function } r(x) = \frac{f(x)}{\bar{F}(x)} = \frac{cx^{c-1}}{b^c}, \quad x \geq 0.$$

For  $c < 1$ ,  $r(x)$  is a decreasing function of  $x$ ;

For  $c = 1$ ,  $r(x) = 1/b$ , i.e. a constant;

For  $c > 1$ ,  $r(x)$  is an increasing function of  $x$ .