### **Functions of Random Variables**

Let's consider a continuous r.v., Y, which is a differentiable, increasing function of a second continuous r.v., X:

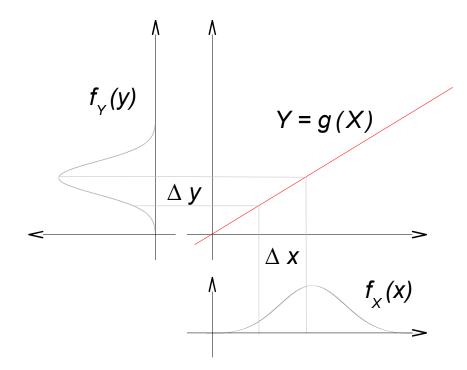
$$Y = g(X).$$

Because g is differentiable and increasing, g' and  $g^{-1}$  are guaranteed to exist. Because g maps all  $x \le s \le x + \Delta x$  to  $y \le t \le y + \Delta y$ :

$$\int_{x}^{x+\Delta x} f_X(s) ds = \int_{y}^{y+\Delta y} f_Y(t) dt.$$

It follows that for small  $\Delta x$ :

$$f_Y(y)\Delta y \approx f_X(x)\Delta x.$$



Dividing by  $\Delta y$ , we get an approximate expression for  $f_Y$  in terms of  $f_X$ :

$$f_Y(y) \approx f_X(x) \frac{\Delta x}{\Delta y}.$$

This is exact in the limit  $\Delta x \rightarrow 0$ :

$$egin{aligned} f_Y(y) &= \lim_{\Delta x o 0} f_X(x) rac{\Delta x}{\Delta y} \ &= \lim_{\Delta x o 0} f_X(x) rac{1}{\Delta y/\Delta x} \ &= \lim_{\Delta x o 0} rac{f_X(x)}{\Delta y/\Delta x}. \end{aligned}$$

From calculus we know that:

 $\lim_{\Delta x \to 0} \frac{\Delta y}{\Delta x} = \lim_{\Delta x \to 0} \frac{g(x + \Delta x) - g(x)}{\Delta x} = g'(x).$ Consequently

$$f_Y(y) = \frac{f_X(x)}{g'(x)}.$$

Substituting  $g^{-1}(y)$  for *x*:

$$f_Y(y) = \frac{f_X(g^{-1}(y))}{g'(g^{-1}(y))}$$

yields an expression for  $f_Y$  in terms of g',  $g^{-1}$ , and  $f_X$ .

# Linear Example

Consider a continuous r.v., *Y*, which is a linear function of a continuous r.v., *X*. Specifically, Y = aX + b. It follows that

$$g(x) = ax + b$$
  

$$g'(x) = a$$
  

$$g^{-1}(y) = (y-b)/a.$$

Substituting the above into

$$f_Y(y) = \frac{f_X(g^{-1}(y))}{g'(g^{-1}(y))}$$

yields

$$f_Y(y) = \frac{1}{a} f_X\left(\frac{y-b}{a}\right).$$

Linear Example (contd.)

Let 
$$f_X(x) = \frac{1}{\sqrt{2\pi}} e^{-(x-\mu)^2/2}$$
, then  
 $f_Y(y) = \frac{1}{a\sqrt{2\pi}} e^{-\left(\frac{y-b}{a}-\mu\right)^2/2}.$ 

# Linear Example (contd.)

Unfortunately, there is a problem. When a = -1, the function g is not increasing (it is decreasing). Consequently,

$$-f_Y(y)\Delta y \approx f_X(x)\Delta x.$$

It follows that for decreasing functions,

$$f_Y(y) = \frac{f_X(g^{-1}(y))}{-g'(g^{-1}(y))}.$$

However, we can derive a function which is correct in both cases by replacing g'(.)with |g'(.)| in the expression relating  $f_Y(.)$ and  $f_X(.)$ :

$$f_Y(y) = \frac{f_X(g^{-1}(y))}{|g'(g^{-1}(y))|}.$$

## Quadratic Example

Consider a continuous r.v., *Y*, which is a quadratic function of a continuous r.v., *X*. Specifically,  $Y = X^2$ . It follows that

$$g(x) = x^2$$
  

$$g'(x) = 2x$$
  

$$g^{-1}(y) = \sqrt{y}.$$

Substituting the above into

$$f_Y(y) = \frac{f_X(g^{-1}(y))}{|g'(g^{-1}(y))|}$$

yields

$$f_Y(y) = \frac{f_X(\sqrt{y})}{|2\sqrt{y}|}.$$

Quadratic Example (contd.)

Let 
$$f_X(x) = \frac{1}{\sqrt{2\pi}} e^{-x^2/2}$$
, then  
 $f_Y(y) = \frac{1}{\sqrt{2\pi}} \frac{e^{-y/2}}{|2\sqrt{y}|}.$ 

Unfortunately, there is a problem:

$$\int_0^\infty f_Y(y) dy = \int_0^\infty \frac{1}{\sqrt{2\pi}} \frac{e^{-y/2}}{|2\sqrt{y}|} dy = \frac{1}{2} \neq 1.$$

Can anyone see the mistake?

### Quadratic Example (contd.)

The mistake is that two different values of *Y* satisfy  $Y = X^2$ :

$$g^{-1}(y) = \pm \sqrt{y}.$$

We decided to use the positive square root arbitrarily and ignored the negative square root. Hence the factor of two error. In general, if a function does not have a unique inverse, we must sum over all possible inverse values:

$$f_Y(y) = \sum_{i=1}^n \frac{f_X(g_i^{-1}(y))}{|g'(g_i^{-1}(y))|}$$

Quadratic Example (contd.)

Let  $g_1^{-1}(y) = \sqrt{y}$  and  $g_2^{-1}(y) = -\sqrt{y}$ , then

$$f_Y(y) = \frac{f_X(g_1^{-1}(y))}{|g'(g_1^{-1}(y))|} + \frac{f_X(g_2^{-1}(y))}{|g'(g_2^{-1}(y))|} \\ = \frac{1}{\sqrt{2\pi}} \frac{e^{-y/2}}{\sqrt{y}}.$$

This is an example of the  $\chi^2$  distribution.

### Kinetic Energy

Recall from physics, that the *kinetic energy*, *K*, of a moving body is given by

$$K = \frac{1}{2}mV^2$$

where *m* is mass and *V* is velocity. Let *V* be a normally distributed random variable with mean,  $\mu$ , and variance,  $\sigma^2$ :

$$f_V(v) = \frac{1}{\sigma\sqrt{2\pi}}e^{-(v-\mu)^2/2\sigma^2}.$$

We would like to compute  $f_K(k)$ , the p.d.f for the continuous random variable, *K*.

# Kinetic Energy (contd.)

We start by computing, g',  $g_1^{-1}$ , and  $g_2^{-1}$ :

$$g(v) = rac{1}{2}mv^2$$
  
 $g'(v) = mv$   
 $g_1^{-1}(k) = \sqrt{2k/m}$   
 $g_2^{-1}(k) = -\sqrt{2k/m}.$ 

Substituting  $f_V(v)$  and the above into

$$f_K(k) = \frac{f_V(g_1^{-1}(k))}{|g'(g_1^{-1}(k))|} + \frac{f_V(g_2^{-1}(k))}{|g'(g_2^{-1}(k))|}$$

yields

$$f_{K}(k) = \frac{e^{-\left(\sqrt{2k/m}-\mu\right)^{2}/2\sigma^{2}} + e^{-\left(-\sqrt{2k/m}-\mu\right)^{2}/2\sigma^{2}}}{\sigma\sqrt{2\pi} m\sqrt{2k/m}}$$
$$= \frac{e^{-\left(\sqrt{2k/m}-\mu\right)^{2}/2\sigma^{2}} + e^{-\left(\sqrt{2k/m}+\mu\right)^{2}/2\sigma^{2}}}{2\sigma\sqrt{\pi k m}}.$$