**Theorem** [UNDER CONSTRUCTION!] If  $X_i \sim N(\mu, \sigma^2)$ , i = 1, 2, ..., n are mutually independent and identically distributed random variables, then  $Y = \sum_{i=1}^{n} X_i^2 / \sigma^2$  has the noncentral chi-square distribution.

**Proof** [UNDER CONSTRUCTION!] Let  $X_i, i = 1, 2, ..., n$  have the  $N(\mu, \sigma^2)$  distribution with probability density function

$$f_{X_i}(x) = \frac{1}{\sqrt{2\pi\sigma}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2} \qquad -\infty < x < \infty$$

The transformation  $Y_i = g(X_i) = X_i/\sigma$  is a 1–1 transformation from  $\mathcal{X} = \{x \mid -\infty < x < \infty\}$  to  $\mathcal{Y} = \{y \mid -\infty < y < \infty\}$  with inverse  $X_i = g^{-1}(Y_i) = \sigma Y_i$  and Jacobian

$$\frac{dX_i}{dY_i} = \sigma.$$

Using the transformation technique, the probability density function of  $Y_i$  is

$$f_{Y_i}(y) = f_{X_i}\left(g^{-1}(y)\right) \left| \frac{dx}{dy} \right|$$
  
=  $\frac{1}{\sqrt{2\pi\sigma}} e^{-\frac{1}{2}\left(\frac{\sigma y - \mu}{\sigma}\right)^2} |\sigma|$   
=  $\frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(y - \mu/\sigma)^2} \qquad -\infty < y < \infty.$ 

Therefore,  $Y_i \sim N(\mu/\sigma, 1)$ . Let  $V_i = h(Y_i) = Y_i^2$ . This is a 2–1 transformation from  $\mathcal{Y} = \{y \mid -\infty < y < \infty\}$  to  $\mathcal{Y} = \{v \mid v > 0\}$ . The domain of the transformation can be divided into  $\mathcal{Y}_1 = \{y \mid y \leq 0\}$  and  $\mathcal{Y}_2 = \{y \mid y > 0\}$ , such that the mapping from  $\mathcal{Y}_1$  to  $\mathcal{Y}$  and  $\mathcal{Y}_2$  to  $\mathcal{Y}$  are each 1–1. The inverse functions are  $Y_i = h_1^{-1}(V_i) = -\sqrt{V_i}$  and  $Y_i = h_2^{-1}(V_i) = \sqrt{V_i}$ , and the Jacobians are

$$J_1 = \frac{dY_i}{dV_i} = -\frac{1}{2\sqrt{V_i}}$$

and

$$J_2 = \frac{dY_i}{dV_i} = \frac{1}{2\sqrt{V_i}}$$

Using the transformation technique, the probability density function of  $V_i$  is

$$\begin{aligned} f_{V_i}(v) &= f_{Y_i}\left(h_1^{-1}(v)\right) |J_1| + f_{Y_i}\left(h_2^{-1}(v)\right) |J_2| \\ &= \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}\left(-\sqrt{v}-\mu/\sigma\right)^2} \left|-\frac{1}{2\sqrt{v}}\right| + \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}\left(\sqrt{v}-\mu/\sigma\right)^2} \left|\frac{1}{2\sqrt{v}}\right| \\ &= \frac{1}{2\sqrt{2\pi v}} \left(e^{-\frac{1}{2}\left(-\sqrt{v}-\mu/\sigma\right)^2} + e^{-\frac{1}{2}\left(\sqrt{v}-\mu/\sigma\right)^2}\right) \qquad v_i > 0 \end{aligned}$$

This proof is not complete. The result is given on page 75 of Forbes, Evans, Hastings, and Peacock (Statistical Distributions, fourth edition, John Wiley and Sons, 2011).