

If X is normal, mean 0, s.d. 1, then $W = -X$ is normal, mean 0, s.d. 1

Proof:



Find $F_W(w_0)$: $F_W(w_0) = P(W \leq w_0) = P(-X \leq w_0) = P(X \geq -w_0) =$
(Way 1)

$$= 1 - P(X \leq -w_0) = 1 - \int_{-\infty}^{-w_0} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx = \left. \begin{array}{l} \text{let } t = -x \\ dt = -dx \\ dx = (-1)dt \\ \text{if } x = -\infty, t = +\infty \\ \text{if } x = -w_0, t = +w_0 \end{array} \right\}$$

$$= 1 - \int_{+\infty}^{+w_0=t} \frac{1}{\sqrt{2\pi}} e^{-\frac{(-t)^2}{2}} (-1) dt$$

$$= 1 + \int_{\infty}^{w_0} \frac{1}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt = 1 - \int_{w_0}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt \quad (\text{since } \int_a^b = -\int_b^a)$$

$$= \int_{-\infty}^{w_0} \frac{1}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt$$

so $F_W(w) = \int_{-\infty}^w \frac{1}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt$, $f_W(w) = \frac{1}{\sqrt{2\pi}} e^{-\frac{w^2}{2}}$, and W is Normal, mean 0, s.d. 1

(Way 2, slicker:)

$$P(X \geq -w_0) = \int_{-w_0}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx = \int_{+w_0=t}^{-\infty=t} \frac{1}{\sqrt{2\pi}} e^{-\frac{(-t)^2}{2}} (-1) dt \left. \begin{array}{l} \text{let } t = -x \\ dx = (-1)dt \\ \text{if } x = \infty, t = -\infty \\ \text{if } x = -w_0, t = +w_0 \end{array} \right\}$$

$$= - \int_{w_0}^{-\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt = + \int_{-\infty}^{w_0} \frac{1}{\sqrt{2\pi}} e^{-\frac{t^2}{2}} dt, \text{ finish as before.}$$

With coverage from the simple case, Prove: If X is normal μ, σ and

$Y = aX + b$, then Y is normal, mean $(a\mu + b)$, s.d. $|a|\sigma$, in the case $a < 0$.

Proof: Since $a < 0$, $-|a| = a$, switch notations as convenient.

$$F_Y(y_0) = P(Y \leq y_0) = P(aX + b \leq y_0) = P(X \geq \frac{y_0 - b}{-|a|}) = \int_{\frac{y_0 - b}{-|a|}}^{\infty} \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2}(\frac{x - \mu}{\sigma})^2} dx =$$

Letting $t = aX + b$, $x = \frac{t - b}{-|a|}$, $dx = \frac{1}{-|a|} dt$

If $x = \infty$, $t = -\infty$. If $x = \frac{y_0 - b}{-|a|}$, $t = a(\frac{y_0 - b}{-|a|}) + b = y_0$

$$= \int_{y_0=t}^{-\infty=t} \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2}(\frac{x - \mu}{\sigma})^2} (\frac{1}{-|a|}) dt$$

$$= \int_{-\infty}^{y_0} \frac{1}{\sqrt{2\pi}|a|\sigma} e^{-\frac{1}{2}(\frac{t - (a\mu + b)}{|a|\sigma})^2} dt = \int_{-\infty}^{y_0} \frac{1}{\sqrt{2\pi}|a|\sigma} e^{-\frac{1}{2}(\frac{t - (a\mu + b)}{|a|\sigma})^2} dt$$

so Y is normal, mean $(a\mu + b)$, s.d. $|a|\sigma$

(Way 3) Using Fund. theorem derivative,

$$F_Y(y_0) = P(X > \frac{y_0 - b}{-|a|}) = \int_{\frac{y_0 - b}{-|a|}}^{\infty} \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}(\frac{x - \mu}{\sigma})^2} dx.$$

$$F_Y'(y_0) = \frac{d}{dy} \left[- \int_{\frac{y_0 - b}{-|a|}}^{\infty} \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}(\frac{x - \mu}{\sigma})^2} dx \right] = - \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}(\frac{(y_0 - b)}{-|a|} - \mu)^2} \cdot \frac{d}{dy} \left(\frac{y_0 - b}{-|a|} \right)$$

$$= \frac{1}{|a|\sqrt{2\pi}} \cdot e^{-\frac{1}{2}(\frac{y_0 - (a\mu + b)}{|a|\sigma})^2} = f_Y(y)$$