## Integrating the Gaussian.

For each real x we let

$$\phi(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}$$

and we let

$$\Phi(x) = \int_{-infty}^{x} \phi(w) \, dw.$$

 $\phi$  is called the standard Gaussian (I think) and  $\Phi$  is called the standard error function.

Theorem. We have

$$\int_{-\infty}^{\infty} \phi(x) \, dx = 1.$$

**One of many proofs.** We have

$$\begin{split} \left(\int_{-\infty}^{\infty} e^{-\frac{x^2}{2}} dx\right)^2 &= \left(\int_{-\infty}^{\infty} e^{-\frac{x^2}{2}} dx\right) \left(\int_{-\infty}^{\infty} e^{-\frac{y^2}{2}} dy\right) \\ &= \int_{-\infty}^{\infty} \left(\int_{-\infty}^{\infty} e^{-\frac{x^2}{2}} dx\right) e^{-\frac{y^2}{2}} dy \\ &= \int_{-\infty}^{\infty} \left(\int_{-\infty}^{\infty} e^{-\frac{x^2}{2}} e^{-\frac{y^2}{2}} dx\right) dy \\ &= \int \int_{\mathbf{R}^2} e^{-\frac{x^2}{2}} e^{-\frac{y^2}{2}} dx dy \\ &= \int \int_{\mathbf{R}^2} e^{-\frac{x^2+y^2}{2}} dx dy \\ &= \int \int_{(0,\infty)\times(0,2\pi)} e^{-\frac{r^2}{2}} r \, dr d\theta \\ &= \int_{0}^{\infty} \left(\int_{0}^{2\pi} e^{-\frac{r^2}{2}} r \, dr\right) dr \\ &= 2\pi \int_{0}^{\infty} e^{-\frac{r^2}{2}} r \, dr \\ &= 2\pi \left[-e^{\frac{-r^2}{2}}\right]_{r=0}^{r=\infty} \\ &= 2\pi. \end{split}$$

**Definition.** We say the random variable X is **standard normal** if  $F_X = \Phi$ . This implies that X is continuous with  $f_X = \phi$ .

**Theorem.** Suppose X is standard normal. Then X has mean 0 and variance 1.

**Proof.** We have

$$E(X) = \frac{1}{\sqrt{2\pi}} \int_{-infty}^{\infty} x e^{-\frac{x^2}{2}} dx = 0$$

because  $x \mapsto xe^{-\frac{x^2}{2}}$  is odd. Also,

$$\sqrt{2\pi}E(X^2) = \int_{-\infty}^{\infty} x^2 e^{-\frac{x^2}{2}} dx$$
$$= -\int_{-\infty}^{\infty} x d\left(e^{-\frac{x^2}{2}}\right)$$
$$= \int_{-\infty}^{\infty} e^{-\frac{x^2}{2}} dx$$
$$= \sqrt{2\pi}.$$

**Definition.** Suppose  $-\infty < \mu < \infty$  and  $0 < \sigma < \infty$ . We say the random variable X is normally distributed with parameter  $\mu$  and  $\sigma^2$  or normally distributed with mean  $\mu$  and variance  $\sigma$  if

$$\frac{X-\mu}{\sigma}$$

is standard normal; if this is the case we immediately infer that the mean of X is  $\mu$  and the variance of X is  $\sigma^2$ .