

## THE NORMAL DISTRIBUTIONS $\mathcal{N}(\mu, \sigma)$

Let  $\mu$  and  $\sigma$  be real numbers, with  $\sigma > 0$ , and suppose  $X$  is a random variable with mean  $\mu$  and standard deviation  $\sigma$ . To say that  $X$  has the  $\mathcal{N}(\mu, \sigma)$  distribution is the same thing as saying that its standardized version

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

has the standard normal distribution  $\mathcal{N}(0, 1)$ , that is,  $Z$  has the distribution given by the probability density function

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

Note, we have  $X = \mu + \sigma Z$ , and for any numbers  $a, b$  with  $a \leq b$ , each of the following conditions is equivalent:

$$\begin{aligned} a &\leq Z && \leq b , \\ a\sigma &\leq \sigma Z && \leq b\sigma , \\ \mu + a\sigma &\leq \mu + \sigma Z && \leq \mu + b\sigma , \\ \mu + a\sigma &\leq X && \leq \mu + b\sigma , \\ a\sigma &\leq X - \mu && \leq b\sigma . \end{aligned}$$

Because these conditions define the same event, they have the same probability, in particular

$$\begin{aligned} \text{Prob}(a \leq Z \leq b) &= \text{Prob}(a\sigma \leq X - \mu \leq b\sigma) = \text{Prob}(\mu + a\sigma \leq X \leq \mu + b\sigma) \\ &*** \text{Prob}(a \leq Z \leq b) = \text{Prob}(\mu + a\sigma \leq X \leq \mu + b\sigma) *** . \end{aligned}$$

For example, if  $\mu = 7$  and  $\sigma = 3$ , and we take  $a = -2, b = 2$ , then

$$\begin{aligned} \text{Prob}(-2 \leq Z \leq 2) &= \text{Prob}(-2(3) \leq X - 7 \leq 2(3)) \\ &= \text{Prob}(7 - 2(3) \leq X \leq 7 + 2(3)) = \text{Prob}(1 \leq X \leq 13) . \end{aligned}$$

SO: the distribution  $\mathcal{N}(\mu, \sigma)$  looks just like  $\mathcal{N}(0, 1)$ , except it is recentered at  $\mu$  and rescaled by  $\sigma$ .

If the standardized version  $Z$  of  $X$  is approximately  $\mathcal{N}(0, 1)$ , then “=” in the last four equations becomes “approximately equals”.

It's important to understand that for the normal distribution, the area under the curve (the graph of the p.d.f.) outside  $[-t, t]$  falls off very rapidly as  $t$  gets large.

Let  $\Phi$  denote the cumulative distribution function for the standard normal distribution. I.e., if  $Z \sim \mathcal{N}(0, 1)$ , then  $\text{Prob}(Z \leq t) = \Phi(t)$ . Then (using = to mean = to the indicated decimal places (where the rightmost digit might be off by at most 1): when  $Z$  is standard normal, here is a table:

$k$	$\text{Prob}( Z  > k)$
1	.32
2	.046
3	.002 7
4	.000 063
5	.000 000 57
6	.000 000 002 0
7	.000 000 000 002 6
8	.000 000 000 000 001 2
9	.000 000 000 000 000 000 23
10	.000 000 000 000 000 000 000 015 = $(1.5)10^{-23}$

The distribution of ANY normal random variable about its mean, measured in units of standard deviations, gives the exact same numbers.

For example, if  $X$  is a normally distributed random variable with mean  $\mu$  and standard deviation  $\sigma$ , then  
 $\text{Prob}(|X - \mu| > 2\sigma) = .046$  and  $\text{Prob}(|X - \mu| > 4\sigma) = .000 063$ .