

Normal probability density functions.

Recall:

(1) An rv X whose range is some interval (c,d) of real numbers is a continuous rv.

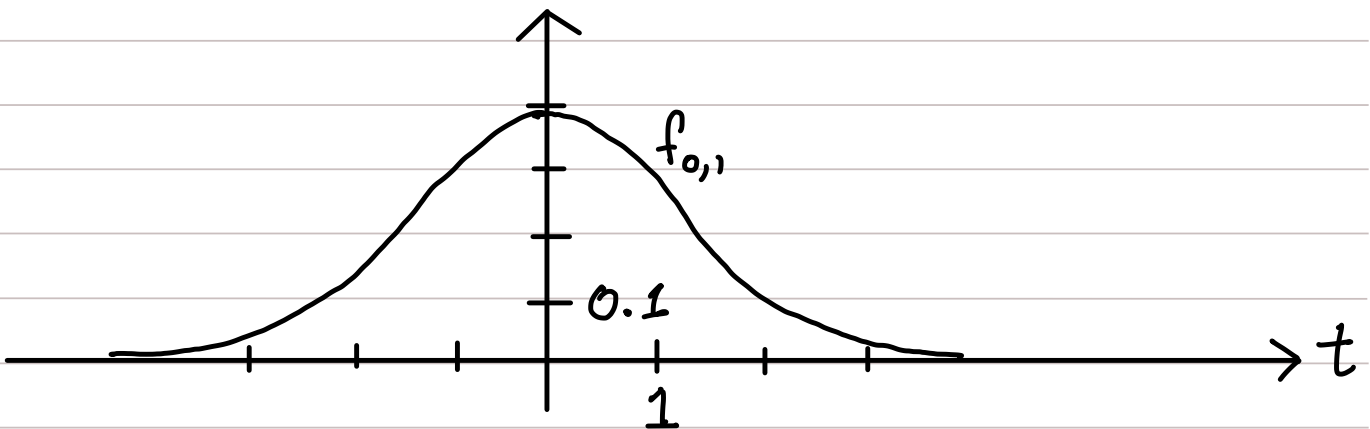
(2) The probability density function, or pdf, for such an X is a function f on (c,d) satisfying

$$P(a < X < b) = \int_a^b f(x) dx.$$

(A) The standard normal pdf.

$$\text{Let } f_{0,1}(x) = \frac{1}{\sqrt{2\pi}} e^{-x^2/2}.$$

Then $f_{0,1}$ is a pdf (see H(10)10), with mean $\mu = 0$ and std dev $\sigma = 1$, and domain $(-\infty, \infty)$.



If X is a population or an rv with pdf $f_{0,1}$, we say " X is standard normal" or " X is $N(0,1)$ ".

Then of course, for $-\infty < a \leq b < \infty$,

$$P(a < X < b) = \int_a^b f_{0,1}(x) dx = \frac{1}{\sqrt{2\pi}} \int_a^b e^{-x^2/2} dx.$$

Note:

$f_{0,1}$ has no simple antiderivative, so in general, to compute standard normal probabilities, we need Riemann sums.

For example, it's known that, if $X = N(0, 1)$, then

(i) $P(-1 < X < 1) = 0.683$. That is: in a standard normal distribution, 68.3% of the data is within one std. dev. of the mean.

Similarly,

$$P(-2 < X < 2) = 0.955,$$

$$P(-3 < X < 3) = 0.997.$$

(ii) $P(-1.96 < X < 1.96) = 0.95$. In a standard normal distribution, 95% of the data is within 1.96 std devs. of the mean.

Similarly,

$$P(-2.33 < X < 2.33) = 0.98,$$

$$P(-2.58 < X < 2.58) = 0.99.$$

(B) Other normal pdf's. If

(3)

$$f_{\mu,\sigma}(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

then one checks that $f_{\mu,\sigma}$ is a pdf, with mean μ and std dev σ . (And $f_{\mu,\sigma}$ is normal - it has the same basic shape as $f_{0,1}$.)

If Z is an rv with pdf $f_{\mu,\sigma}$, we say " Z is $N(\mu,\sigma)$ ".

Useful idea: to know $N(\mu,\sigma)$ probabilities, it's enough to know $N(0,1)$ probabilities.

Here's why:

NISNID (normal is standard normal in disguise)
fact:

If Z is $N(\mu,\sigma)$, then

$$X = \frac{Z - \mu}{\sigma} \text{ is } N(0,1).$$

[Proof omitted.]

Here's how we use this fact:

Examples.

(1) Suppose Z is $N(2,0.5)$. Find $P(1.02 < Z < 2.98)$.

Solution.

"standardize" Z (subtract μ , then divide by σ)

$$P(1.02 < Z < 2.98) \stackrel{\downarrow}{=} P\left(\frac{1.02-2}{0.5} < \frac{Z-2}{0.5} < \frac{2.98-2}{0.5}\right)$$

$$= P\left(-1.96 < \frac{Z-2}{0.5} < 1.96\right) = 0.95 = 95\%.$$

by NISNID fact
and by A(ii) above

(2) Suppose Y is $N(-1, 3)$. Find $P(-4 < Y < 2)$.

Solution.

$$\begin{aligned} P(-4 < Y < 2) &= P\left(\frac{-4 - (-1)}{3} < \frac{Y - (-1)}{3} < \frac{2 - (-1)}{3}\right) \\ &= P\left(-1 < \frac{Y - (-1)}{3} < 1\right) \end{aligned}$$

by NSNID
fact and by
A(i) above

$$= 0.683 = 68.3\%.$$

(3) Suppose Z is $N(\mu, \sigma)$. Find the proportion of data in Z that's within 3 std devs of the mean.

Solution

We're looking for:

$$\begin{aligned} P(\mu - 3\sigma < Z < \mu + 3\sigma) \\ &= P\left(\frac{\mu - 3\sigma - \mu}{\sigma} < \frac{Z - \mu}{\sigma} < \frac{\mu + 3\sigma - \mu}{\sigma}\right) \\ &= P(-3 < \frac{Z - \mu}{\sigma} < 3) = 99.7\%, \end{aligned}$$

by A(i) above.