

CONTINUOUS RANDOM VARIABLES

The normal distribution

The normal distribution

- The normal (Gaussian) distribution is certainly the most commonly used continuous distribution as a model in statistics.
- Many real-world data sets seem to follow the normal (“bell shaped”) distribution.
- The normal distribution is defined over the entire real line.



The normal distribution

- Definition 5: A random variable Y has a NORMAL DISTRIBUTION with mean μ and variance σ^2 if its pdf is

$$f(y) = \frac{1}{\sigma\sqrt{2\pi}} e^{-(y-\mu)^2/(2\sigma^2)}, \quad -\infty < y < \infty$$

- Shorthand notation: $Y \sim N(\mu, \sigma^2)$. (Note: must have $\sigma > 0$.)

The normal distribution

- The normal distribution is symmetric about μ .
- It has inflection points $\mu - \sigma$ and $\mu + \sigma$.
- The normal pdf is positive everywhere, but

$$\lim_{y \rightarrow -\infty} f(y) = \lim_{y \rightarrow \infty} f(y) = 0$$

- Because $f(y)$ is a pdf,

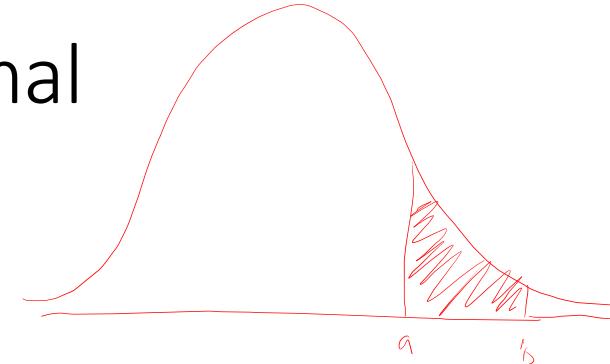
$$\int_{-\infty}^{\infty} \underline{f(y)dy} = \int_{-\infty}^{\infty} \frac{1}{\sigma\sqrt{2\pi}} e^{-(y-\mu)^2/(2\sigma^2)} dy = 1$$

(This is *really hard* to prove!)

Calculating probabilities with the normal distribution

- If $Y \sim N(\mu, \sigma^2)$ then

$$P(a < Y < b) = \int_a^b \frac{1}{\sigma\sqrt{2\pi}} e^{-(y-\mu)^2/(2\sigma^2)} dy$$



- Problem: Nobody knows how to solve this integral!
- Solution: Numerical approximation!
 - Table 4, Appendix 3
 - pnorm() function in R
 - Online calculators
- Assignment: figure out how to use (at least) one of these.
- Useful fact: If $Y \sim N(\mu, \sigma^2)$ then $\frac{Y-\mu}{\sigma} \sim N(0,1)$

$$\underbrace{}_{\text{ }} \quad \overbrace{\frac{Y-\mu}{\sigma}}^{\text{ }} \quad \underbrace{}_{\text{ }} \quad \overbrace{}^{\text{ }} \quad \overbrace{}^{\text{ }}$$

The standard normal distribution

- The parameters of the $N(\mu, \sigma^2)$ distribution are μ and σ^2 .
- If a normal random variable Z has $\mu = 0$ and $\sigma^2 = 1$ (i.e., $Z \sim N(0,1)$) then we say that Z has the standard normal distribution.

Moment generating function of the standard normal distribution

- Theorem 4: If $Z \sim N(0,1)$ then Z has mgf

$$m(t) = E[e^{tZ}] = e^{\frac{1}{2}t^2}$$

$$\begin{aligned}
 m(t) &= E[e^{tZ}] = \int_{-\infty}^{\infty} e^{tz} \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}} dz \\
 &\quad \left[e^{tz} e^{-\frac{z^2}{2}} \right]_{-\infty}^{\infty} = e^{-\frac{t^2}{2}} = e^{-\frac{1}{2}(z-t)^2 + \frac{1}{2}t^2} = e^{-\frac{1}{2}(z-t)^2} e^{\frac{1}{2}t^2} \\
 &= \int_{-\infty}^{\infty} e^{\frac{1}{2}t^2} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(z-t)^2} dz = e^{\frac{1}{2}t^2} \int_{-\infty}^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}(z-t)^2} dz
 \end{aligned}$$

pdf of $N(t, 1)$??

Moment generating function of the normal distribution

- Theorem 5: If $\underline{Y \sim N(\mu, \sigma^2)}$ then Y has mgf

$$m(t) = E[e^{tY}] = \int_{-\infty}^{\infty} e^{ty} \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{1}{2}(y-\mu)^2/\sigma^2} dy$$

[substitute $z = \frac{y-\mu}{\sigma} \Leftrightarrow y = z\sigma + \mu$ $dy = \sigma dz$]

$$= \int_{-\infty}^{\infty} e^{t(z\sigma + \mu)} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}z^2} \cancel{\sigma dz}$$

mgf of $e^{t(z\sigma + \mu)} = e^{t\mu + \frac{1}{2}t^2\sigma^2}$

$$= e^{t\mu} \int_{-\infty}^{\infty} e^{tz\sigma} \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2}z^2} dz$$

$$= e^{t\mu} \cancel{e^{\frac{1}{2}(t\sigma)^2}} = \boxed{e^{\mu + \frac{1}{2}\sigma^2 t^2}}$$

Mean and variance of the normal distribution

- Theorem 6: If $Y \sim N(\mu, \sigma^2)$ then

$$\underline{E[Y] = \mu} \text{ and } \underline{Var(Y) = \sigma^2}$$

Take $Z \sim N(0,1)$ want to show $E[Z] = 0$

$$E[Z] = \int_{-\infty}^{\infty} z \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}} dz = \frac{1}{\sqrt{2\pi}} \left[\int_{-\infty}^0 z e^{-\frac{z^2}{2}} dz + \frac{1}{2\pi} \int_0^{\infty} z e^{-\frac{z^2}{2}} dz \right]$$

$$= \frac{1}{\sqrt{2\pi}} \left(\cancel{\int_{-\infty}^0 z e^u du} - \cancel{\int_{-\infty}^0 z e^u du} \right) = 0$$

If $Z \sim N(0,1)$ $Y = \sigma Z + \mu \sim N(\mu, \sigma^2)$

$$E[Y] = E[\sigma Z + \mu] = \sigma E[Z] + \mu = \mu$$