

Section 3.5

Gaussian distribution
(also called normal distribution)

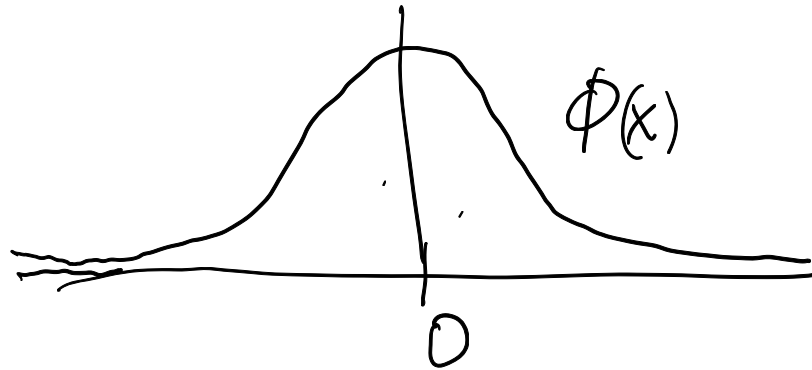
Most simple version of Gaussian dist.

Def: A r.v. Z has the *standard normal dist.* (or standard Gaussian dist.) if it has density function

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

Notation: $Z \sim N(0,1)$.

Plot of pdf:

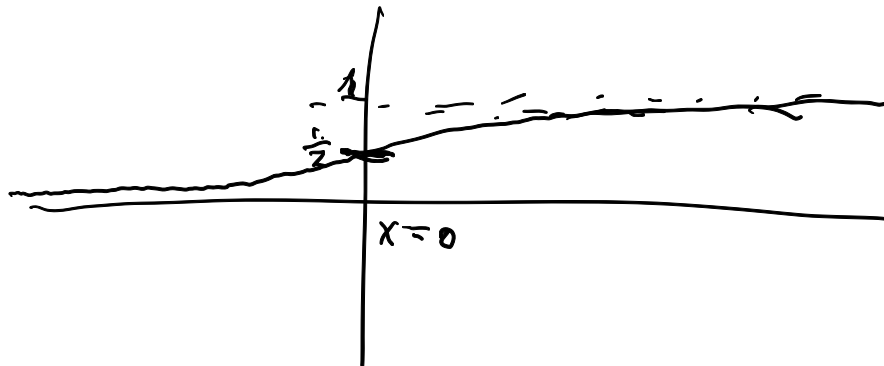


Standard normal

Cdf also gets a special symbol:

$$\Phi(x) = \underbrace{P(Z \leq x)} = \int_{-\infty}^x \phi(s) ds = \underbrace{\frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-\frac{s^2}{2}} ds}_{\text{No closed form soln}}$$

Plot of CDF:



Why the factor of $\frac{1}{\sqrt{2\pi}}$?

Fact: $\int_{-\infty}^{\infty} e^{-\frac{x^2}{2}} dx = \sqrt{2\pi}$ so $\frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-\frac{x^2}{2}} dx = 1$

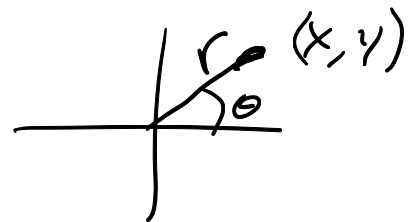
Pf: $\left(\int_{-\infty}^{\infty} e^{-\frac{x^2}{2}} dx \right)^2 = \int_{-\infty}^{\infty} e^{-\frac{x^2}{2}} dx \int_{-\infty}^{\infty} e^{-\frac{y^2}{2}} dy$

$$= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{-\frac{x^2}{2}} \cdot e^{-\frac{y^2}{2}} dx dy = \iint_{\mathbb{R}^2} e^{-\frac{x^2+y^2}{2}} dx dy$$

Change to polar:

$$dx dy = r dr d\theta$$

$$= \int_0^{2\pi} \int_0^{\infty} e^{-\frac{r^2}{2}} \cdot r dr d\theta$$
$$= \int_0^{2\pi} -e^{-\frac{r^2}{2}} \Big|_0^{\infty} d\theta$$



$$= \int_0^{2\pi} (0 - 1) d\theta = -2\pi \quad \checkmark$$

Why called “standard normal”

- **Fact:** If $Z \sim N(0,1)$ has $EZ = 0$, $Var(Z) = 1$

This is why Z is called standard normal.

To get a whole family of (non-standard) normal r.v.s, just consider

$X = \sigma Z + \mu$ for different values of σ and μ .

Let us find the dist of X . First:

$$EX = \underline{E[\sigma Z + \mu]} = \sigma \overset{0}{\cancel{E[Z]}} + \mu = \mu.$$

$$Var(X) = \underline{Var(\sigma Z + \mu)} = \sigma^2 \cancel{Var(Z)} = \sigma^2,$$

Cdf and pdf of $X = \sigma Z + \mu$

$$\text{Cdf: } F(s) = P(X \leq s) = P(\sigma Z + \mu \leq s) = P\left(Z \leq \frac{s - \mu}{\sigma}\right) = \Phi\left(\frac{s - \mu}{\sigma}\right)$$

$$\begin{aligned} \text{Pdf: } f(s) &= \frac{d}{ds} F(s) = \frac{d}{ds} \Phi\left(\frac{s - \mu}{\sigma}\right) = \frac{1}{\sigma} \cdot \phi\left(\frac{s - \mu}{\sigma}\right) \\ &= \frac{1}{\sigma} \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{1}{2} \left(\frac{s - \mu}{\sigma}\right)^2} \end{aligned}$$

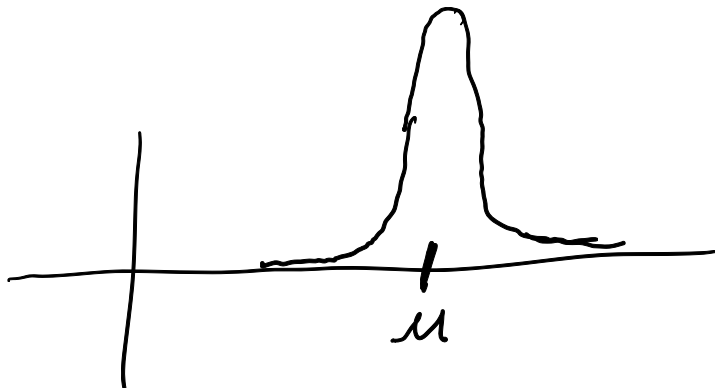
(General) Normal dist (or Gaussian dist)

Def: A r.v. Z has the normal dist. (or Gaussian dist.) if it has density function

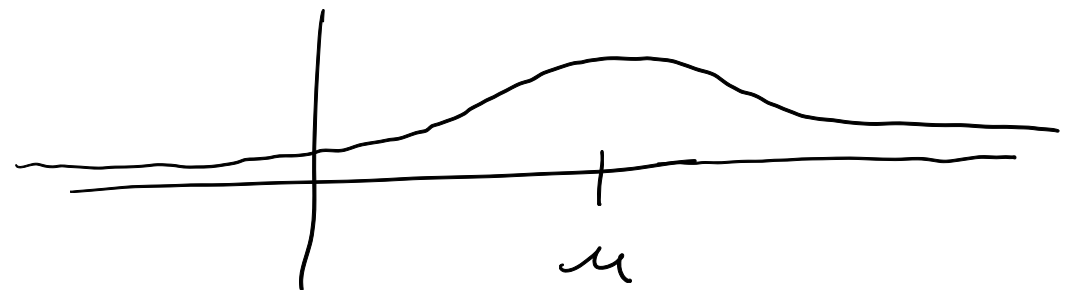
$$f(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}}, \quad \text{for some } \mu \in \mathbb{R}, \sigma > 0.$$

Notation: $Z \sim N(\mu, \sigma^2)$.

Plot: σ small



σ large



Rescaling, translating \rightarrow still Gaussian

Fact: If $X \sim N(\mu, \sigma^2)$, and $Y = aX + b$, then $Y \sim \underline{N}(\underline{a \cdot \mu + b}, \underline{a^2 \cdot \sigma^2})$.

In particular, $\frac{X - \mu}{\sigma} \sim N(0, 1)$

Ex) $X \sim N(-3, 5)$. What is $P(4(X + 5) > 2)$?

$$P(4(X + 5) > 2) = P\left(X > \frac{2}{4} - 5\right) = P\left(X > -\frac{9}{2}\right)$$

$$= P\left(N(-3, 5) > -\frac{9}{2}\right) = P\left(\underbrace{[-3 + \sqrt{5} N(0, 1)]}_{\stackrel{\text{dist}}{=} N(-3, 5)} > -\frac{9}{2}\right)$$

$$= P\left(\sqrt{5} N(0, 1) > \overset{-3/2}{\cancel{3} - \frac{9}{2}}\right)$$

$$= P\left(N(0, 1) > \frac{-3/2}{\sqrt{5}}\right)$$

$$= 1 - P\left(N(0, 1) \leq \frac{-3/2}{\sqrt{5}}\right) = \boxed{1 - \Phi\left(\frac{-3/2}{\sqrt{5}}\right)}$$