

Math 3501 - Probability and Statistics I

3.3 - The normal distribution

the "most important"
probability distribution

Normal distribution

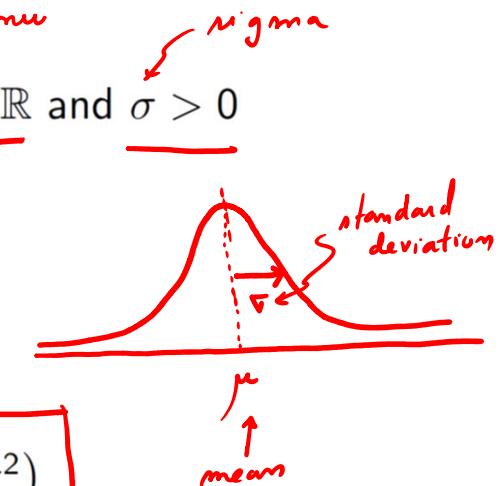
describes how sample averages
behave (under appropriate conditions)

Reason:
Central Limit Theorem
(part of Chp 5)

Normal distribution

A random variable X has a normal distribution with parameters $\mu \in \mathbb{R}$ and $\sigma > 0$ if its pdf is of the form

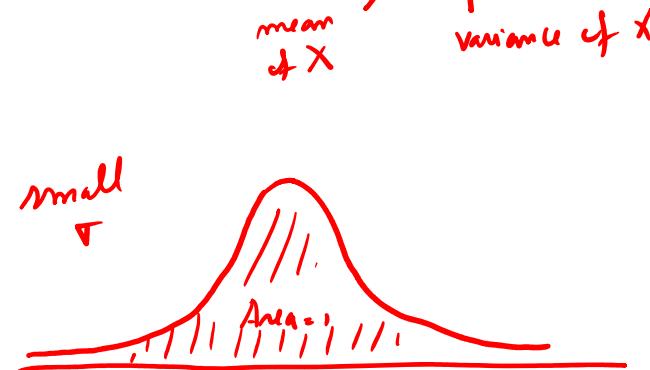
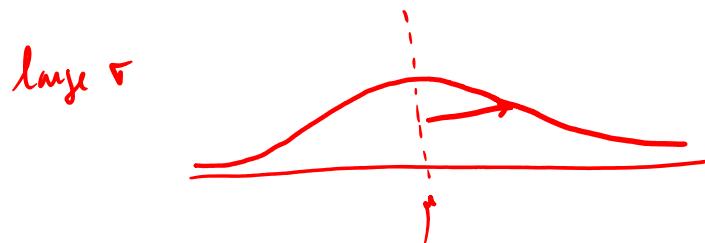
$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} \exp\left[-\frac{(x-\mu)^2}{2\sigma^2}\right], \quad x \in \mathbb{R}.$$



Notation: We say that X is $N(\mu, \sigma^2)$ and denote it as $X \sim N(\mu, \sigma^2)$

Note: It is possible to check that:

- $f(x) > 0$ for all $x \in \mathbb{R}$ ✓
- $\int_{\mathbb{R}} f(x) dx = 1$



Normal distribution mgf, mean and variance

Suppose X has a normal distribution (with parameter μ and σ).

The mgf of X is given by

$$M(t) = E[e^{tX}] = \exp\left(\mu t + \frac{\sigma^2 t^2}{2}\right), \quad x \in \mathbb{R}.$$

$\int_{-\infty}^{\infty} e^{tx} \cdot \frac{1}{\sqrt{2\pi\sigma^2}} \cdot e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx$

Annotations: A red bracket under $E[e^{tX}]$ is labeled "def". A red arrow points from tX to the term $\mu t + \frac{\sigma^2 t^2}{2}$. A red circle around $\sigma^2 t^2$ is labeled "varian 4". A red arrow points from μ to the term $\mu t + \frac{\sigma^2 t^2}{2}$.

Differentiating the mgf $M(t)$, we obtain

$$\begin{aligned} \rightarrow M'(t) &= (\mu + \sigma^2 t) \exp\left(\mu t + \frac{\sigma^2 t^2}{2}\right) \\ \rightarrow M''(t) &= \left[(\mu + \sigma^2 t)^2 + \sigma^2\right] \exp\left(\mu t + \frac{\sigma^2 t^2}{2}\right). \end{aligned}$$

) chain rule
) product +
 + chain
 rule

Evaluating at $t = 0$, we find that

$$\rightarrow E(X) = M'(0) = \mu \quad \leftarrow$$

$$\rightarrow \text{Var}(X) = E(X^2) - [E(X)]^2 = M''(0) - [M'(0)]^2 = (\mu^2 + \sigma^2) - \mu^2 = \sigma^2. \quad \leftarrow$$

Example

Suppose the random variable X has pdf given by

$$f(x) = \frac{1}{\sqrt{32\pi}} \exp\left[-\frac{(x+7)^2}{32}\right], \quad x \in \mathbb{R}.$$

Determine the mean and variance of X .

Observe that f is of the form

$$f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right), \quad x \in \mathbb{R}$$

with

$$\mu = -7$$

$$\text{and } 2\sigma^2 = 32 \Rightarrow \sigma^2 = 16$$

$$\left. \begin{array}{l} X \sim N(-7, 16) \\ \text{and} \\ E[X] = -7 \text{ and } \text{Var}(X) = 16 \end{array} \right\}$$

Example

Suppose the random variable X has mgf given by

$$M(t) = \exp(5t + 12t^2), \quad t \in \mathbb{R}.$$

Determine the mean and variance of X .

Observe that $M(t)$ is of the form

$$M(t) = \exp\left(\mu t + \frac{\sigma^2}{2} t^2\right), \quad t \in \mathbb{R}$$

with

$$\mu = 5$$

and

$$\frac{\sigma^2}{2} = 12$$

$$\text{That is } \mu = 5 \text{ and } \sigma^2 = 24$$

$$\left. \begin{array}{l} X \sim N(5, 24) \\ \text{and} \\ E[X] = 5 \\ \text{Var}(X) = 24 \end{array} \right\} \Rightarrow$$

Standard normal distribution

We say that Z has a standard normal distribution if Z is $\underline{N(0, 1)}$.

The pdf of Z is

lower case phi → $\phi(z)$ → $f(z) = \frac{1}{\sqrt{2\pi}} e^{-z^2/2}$

and its cdf is

Capital phi → $\Phi(z) = P(Z \leq z) = \int_{-\infty}^z \frac{1}{\sqrt{2\pi}} e^{-w^2/2} dw$

Notes:

- It is not possible to evaluate this integral by finding an antiderivative that can be expressed as an elementary function.
- Numerical approximations for integrals of this type have been tabulated.
(Tables Va and Vb in textbook Appendix B)

Φ ← handwritten capital phi

f → pdf

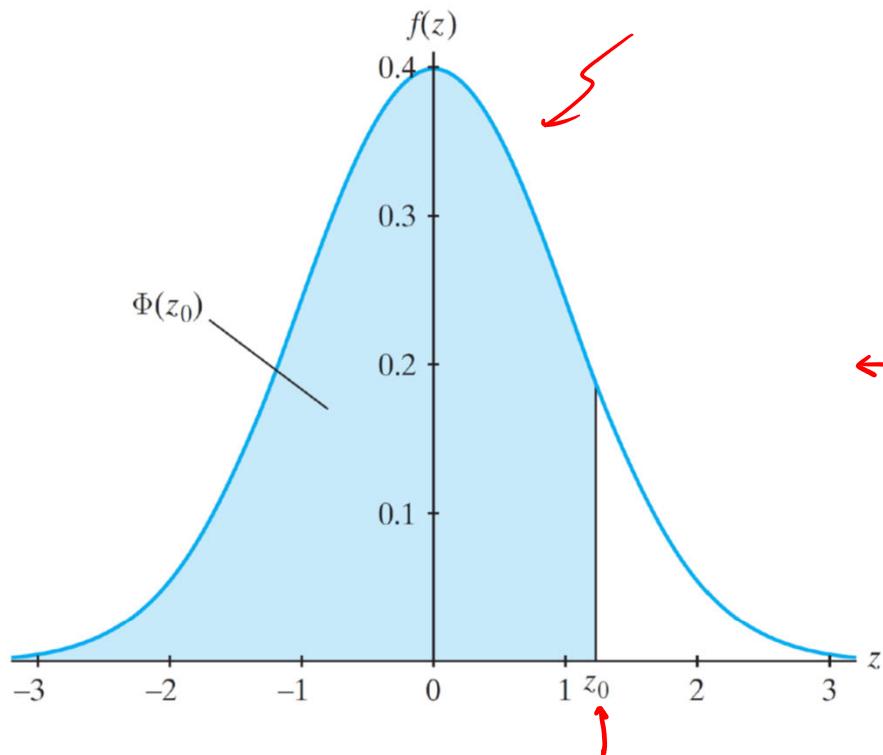
F → cdf

standard normal

ϕ → pdf

Φ → cdf

The graph of the pdf of Z is a bell-shaped curved.



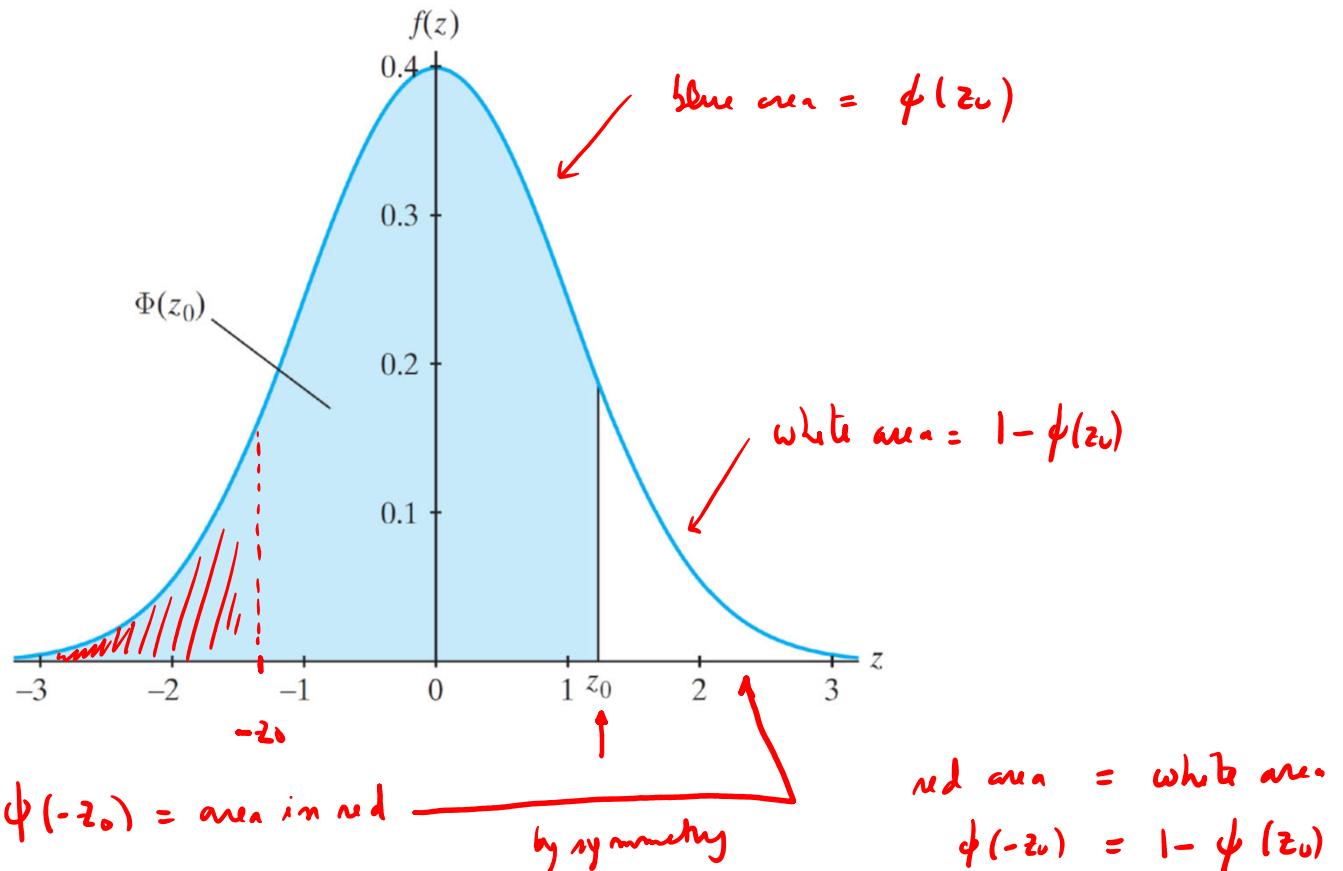
The shaded area equals $\Phi(z_0) = P(Z \leq z_0)$.

cdf of $Z \sim N(0, 1)$

Symmetry guarantees that for all real z , we have

$$\underbrace{\Phi(-z) = P(Z \leq -z)}_{\text{ }} = P(Z > z) = 1 - P(Z \leq z) = \underbrace{1 - \Phi(z)}_{\text{ }}.$$

$$\phi(-z) = 1 - \phi(z)$$



Example

Suppose Z is $N(0, 1)$.

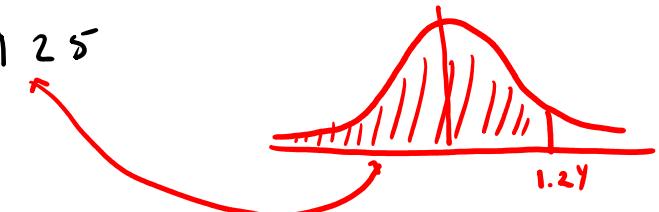
Evaluate each of the following probabilities:

- $P(Z \leq 1.24)$
- $P(1.24 \leq Z \leq 2.37)$
- $P(-2.37 \leq Z \leq -1.24)$

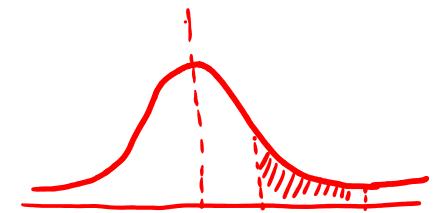
Recall that the cdf of $Z \sim N(0, 1)$

$$\Phi(z) = P(Z \leq z) \leftarrow \text{read from table!}$$

$$P(Z \leq 1.24) = \Phi(1.24) = 0.8925$$



$$P(1.24 < Z \leq 2.37) = P(Z \leq 2.37) - P(Z \leq 1.24)$$



$$= \Phi(2.37) - \Phi(1.24)$$

$$= \phi(2.37) - \phi(1.24)$$

$$= 0.9911 - 0.8925 = \dots \text{use a calculator}$$

$$P(-2.37 \leq Z \leq -1.24) = P(1.24 \leq Z \leq 2.37)$$

= previous item

Pay attention to any symmetry that can be used to simplify the evaluation!!!

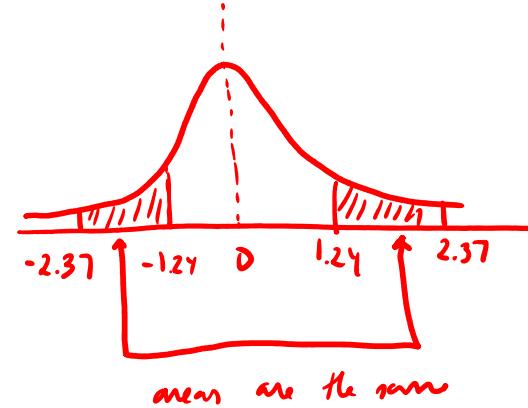
Two more examples:

$$P(Z \leq -1.59) = \phi(-1.59) = 1 - \phi(\underline{1.59}) = 1 - 0.9441 = \text{calculator}$$

table

$$\begin{aligned} P(Z > -0.42) &= 1 - P(Z \leq -0.42) \\ &= 1 - \underbrace{\phi(-0.42)}_{1 - \phi(0.42)} = 1 - \overline{(1 - \phi(0.42))} = 1 - 1 + \phi(0.42) \\ &= \phi(0.42) = 0.6628 \end{aligned}$$

table!



Example

Suppose Z is $N(0, 1)$.

Find constants a and b such that

$$P(Z \leq a) = 0.9147 \quad \text{and} \quad P(Z \geq b) = 0.0526 .$$

$$P(Z \leq a) = 0.9147 \Leftrightarrow \phi(a) = 0.9147 \Leftrightarrow a = 1.37$$

table

$$P(Z \geq b) = 0.0526 \Leftrightarrow 1 - P(Z < b) = 0.0526 \Leftrightarrow 1 - P(Z \leq b) = 0.0526$$

↑
complement

↑
 Z is a continuous r.v

$$\Leftrightarrow 1 - \Phi(b) = 0.0526 \Leftrightarrow$$

$$\Leftrightarrow \Phi(b) = 1 - 0.0526 \Leftrightarrow \phi(b) = 0.9474 \quad \text{table!}$$
$$\Leftrightarrow b = 1.62$$

Percentiles for standard normal distribution

Suppose Z is $N(0, 1)$, and let $\alpha \in (0, 1)$ (usually $\alpha < 0.5$).

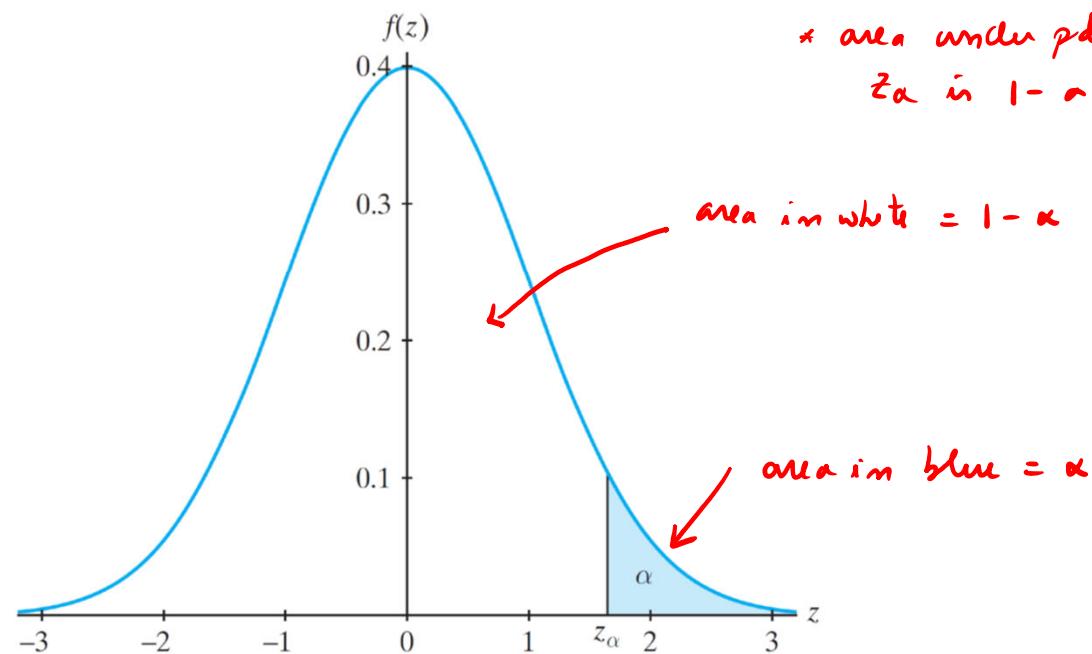
The $100(1 - \alpha)$ th percentile of its distribution is the number z_α such that

$$P[Z \geq z_\alpha] = \alpha,$$

* area under pdf to the right of z_α is α

that is, the probability to the right of z_α is α .

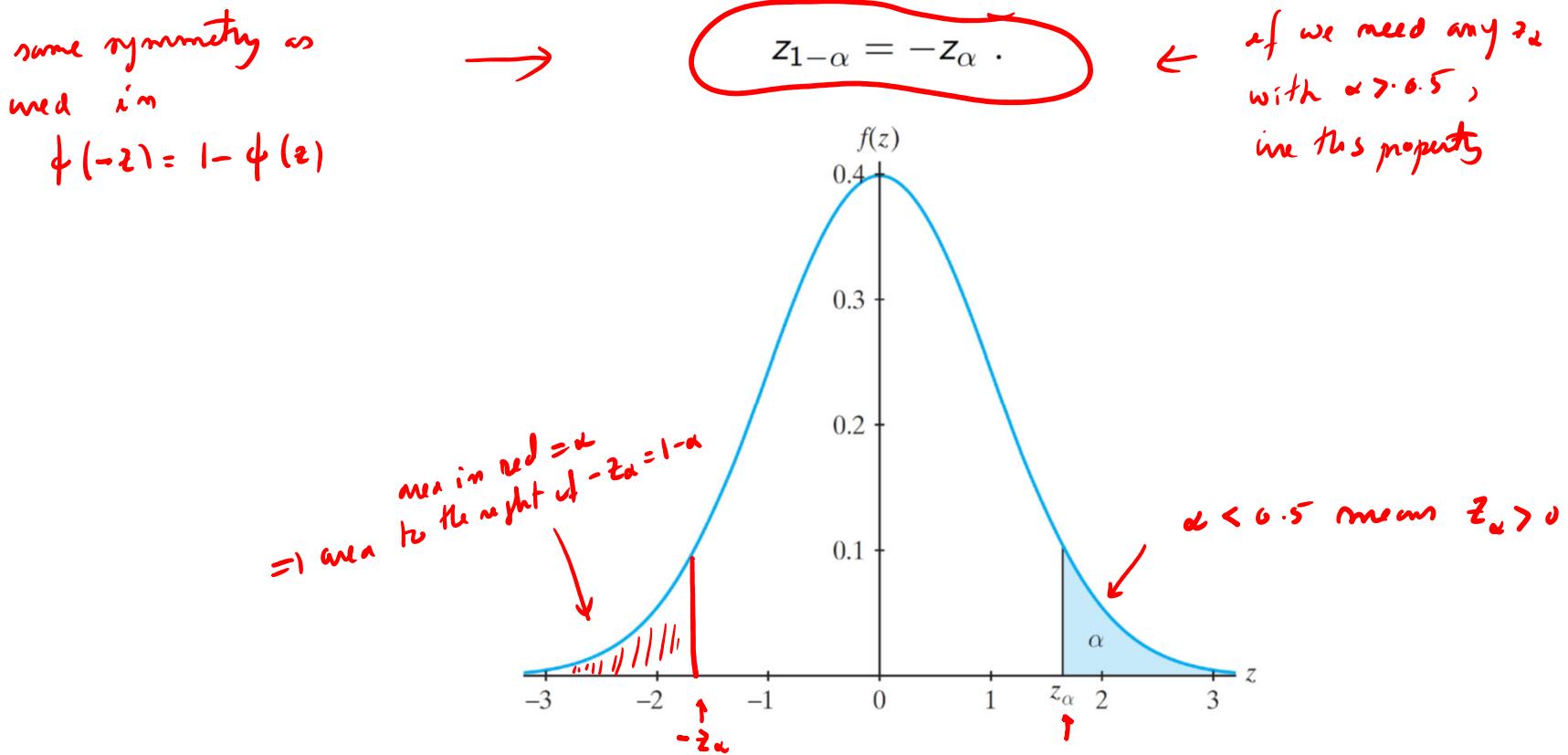
* area under pdf to the left of z_α is $1 - \alpha$



Due to the symmetry of the standard normal pdf, we have

$$P(Z \leq -z_\alpha) = P(Z \geq z_\alpha) = \alpha$$

and, since the subscript of z_α is the right-tail probability, we also have



Example

Find:

$$\bullet z_{0.0125} =$$

$$\bullet z_{0.05} = 1.645 \leftarrow \text{using row for } z_\alpha$$

note that $z_{0.0125} = z_{0.025/2} = 2.240$



look in the row

for $z_{\alpha/2}$ with $\alpha=0.025$

Standard score → allows us to relate any r.v. $X \sim N(\mu, \sigma^2)$ with $Z \sim N(0, 1)$

Theorem

If X is $N(\mu, \sigma^2)$, then

is $N(0, 1)$.

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

Notes:

1. If X is normally distributed, then Z is normally distributed with zero mean and unit variance.
2. Z is often called the standard score associated with X .
3. Standard scores can be used to find probabilities associated with $X \sim N(\mu, \sigma^2)$:

$$P(a \leq X \leq b) = P\left(\frac{a - \mu}{\sigma} \leq \frac{X - \mu}{\sigma} \leq \frac{b - \mu}{\sigma}\right) = \Phi\left(\frac{b - \mu}{\sigma}\right) - \Phi\left(\frac{a - \mu}{\sigma}\right)$$

Example

Suppose X is $N(3, 16)$. $\Rightarrow \mu = 3$ and $\sigma^2 = 16$ (and so $\sigma = 4$)

Evaluate each of the following probabilities:

- $P(4 \leq X \leq 8)$
- $P(0 \leq X \leq 5)$

By the previous theorem, since $X \sim N(\mu, \sigma^2)$, then

$$Z = \frac{X - \mu}{\sigma} = \frac{X - 3}{4} \sim N(0, 1)$$

$$\begin{aligned} P(4 \leq X \leq 8) &= P\left(\frac{4-3}{4} \leq \frac{X-3}{4} \leq \frac{8-3}{4}\right) = P\left(\frac{1}{4} \leq Z \leq \frac{5}{4}\right) \\ &= P(0.25 \leq Z \leq 1.25) = \Phi(1.25) - \Phi(0.25) \\ &= 0.8944 - 0.5987 = \dots \end{aligned}$$

$$\begin{aligned}
 P(0 \leq X \leq 5) &= P\left(\frac{0-3}{4} \leq \underbrace{\frac{X-3}{4}}_Z \leq \frac{5-3}{4}\right) = \\
 &= P\left(-\frac{3}{4} \leq Z \leq \frac{2}{4}\right) = P(-0.75 \leq Z \leq 0.5) \\
 &= \phi(0.5) - \underbrace{\phi(-0.75)}_{1 - \phi(0.75)} \\
 &= \phi(0.5) - (1 - \phi(0.75)) \\
 &= \phi(0.5) - 1 + \phi(0.75) \\
 &= 0.6915 - 1 + 0.7734 = \dots
 \end{aligned}$$

Example

If X is $N(25, 36)$, we find a constant c such that

$$P(|X - 25| \leq c) = 0.9544 .$$

Since $X \sim N(25, 36)$, then $\mu = 25$ and $\sigma = 6$ ($\text{so that } \sigma^2 = 36$)

$$\begin{array}{c} \uparrow \\ \mu \\ \uparrow \\ \sigma^2 \end{array}$$

$$\text{then } Z = \frac{X - \mu}{\sigma} = \frac{X - 25}{6} \sim N(0, 1)$$

$$\text{Observe that } P(|X - 25| \leq c) = 0.9544$$

$$\Rightarrow P(-c \leq X - 25 \leq c) = 0.9544$$

$$\Rightarrow P\left(-\frac{c}{6} \leq \frac{X - 25}{6} \leq \frac{c}{6}\right) = 0.9544$$

Recall the absolute value inequality:

$|x| \leq a$, $a > 0$
is equivalent to
 $-a \leq x \leq a$

$$\Leftrightarrow P\left(-\frac{c}{6} \leq Z \leq \frac{c}{6}\right) = 0.9544$$

$$\Leftrightarrow \phi\left(\frac{c}{6}\right) - \underbrace{\phi\left(-\frac{c}{6}\right)}_{= 1 - \phi\left(\frac{c}{6}\right)} = 0.9544$$

$$\Leftrightarrow \phi\left(\frac{c}{6}\right) - \underbrace{\left(1 - \phi\left(\frac{c}{6}\right)\right)}_{= 2\phi\left(\frac{c}{6}\right) - 1} = 0.9544$$

$$\Leftrightarrow \phi\left(\frac{c}{6}\right) - 1 + \phi\left(\frac{c}{6}\right) = 0.9544$$

$$\Leftrightarrow 2\phi\left(\frac{c}{6}\right) = 1.9544 \quad \Leftrightarrow \phi\left(\frac{c}{6}\right) = \frac{1.9544}{2} = 0.9772$$

$$\Leftrightarrow \phi\left(\frac{c}{6}\right) = 0.9772 \quad \xrightarrow{\text{table}} \quad \frac{c}{6} = 2 \quad \Leftrightarrow \boxed{c = 12}$$

Math 3501 - Probability and Statistics I

4.1 - Bivariate distributions of the discrete type

Joint probability mass function

Definition (Joint probability mass function)

Let X and Y be two random variables defined on a discrete sample space, and let S denote the corresponding two-dimensional space of X and Y , the two random variables of the discrete type.

The joint probability mass function (abbreviated joint pmf) of X and Y , denoted $f(x, y)$, is defined as

$$f(x, y) = P(X = x, Y = y).$$

Properties:

- (a) $0 \leq f(x, y) \leq 1$ for all $(x, y) \in \mathbb{R}^2$
- (b) $\sum_{\substack{(x,y) \in S \\ =}} f(x, y) = 1$
- (c) $P[(X, Y) \in A] = \sum_{(x,y) \in A} f(x, y)$, where $A \subset \mathbb{R}^2$

analogous to 1-dim case (Sec 2.1)

Example

6-faced

Roll a pair of fair dice. Let X denote the smaller and Y the larger outcome on the dice.

Determine the joint pmf of X and Y .

X = r.v. giving the smaller score

Y = r.v. " " " larger score

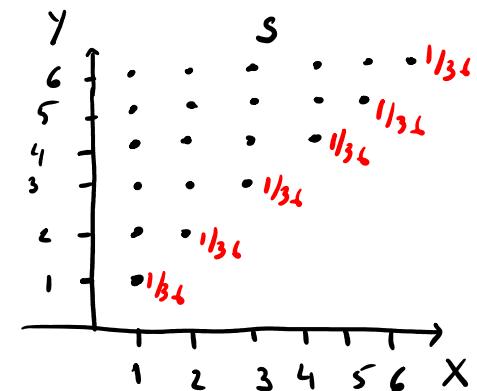
} e.g if scores are $(2,3)$, $X=2$ and $Y=3$
if scores are $(4,4)$, $X=4$ and $Y=4$

The space of X and Y is

$$S = \{(i,j) : i, j \in \{1, 2, \dots, 6\} \text{ and } i \leq j\} \longrightarrow$$

$$P(12,3) = P(X=2, Y=3) = P(\{(2,3), (3,2)\}) = \frac{2}{36} = \frac{1}{18}$$

$$P(4,4) = P(X=4, Y=4) = P(\{(4,4)\}) = \frac{1}{36}$$



$$\text{if } i < j \Rightarrow f(i, j) = \frac{2}{36}$$

$$\text{if } i = j \Rightarrow f(i, j) = \frac{1}{36}$$

$$f(x, y) = \begin{cases} \frac{2}{36} & \text{if } i, j \in \{1, \dots, 6\} \text{ and } i < j \\ \frac{1}{36} & \text{if } i, j \in \{1, \dots, 6\} \text{ and } i = j \end{cases}$$