

Probability and Statistics

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The Normal Distribution

Topics

- 1 The Normal Distribution
 - Introduction
 - The Standard Normal Distribution
 - Normal Distribution Probabilities
 - Percentiles of the Normal Distribution and the z_α Notation
 - The Normal Approximation to the Binomial

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The Normal Distribution

Objectives

Objectives:

- Recognize normal random variables.
- Use the normal distribution to find probabilities.
- Compute and interpret standardized values (z -scores).
- State the Empirical Rule.
- Find percentiles of the normal distribution.
- Use the normal distribution to approximate binomial probabilities.

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The Normal Distribution

Introduction
The Standard Normal Distribution
Normal Distribution Probabilities
Percentiles of the Normal Distribution and the z_α Notation
The Normal Approximation to the Binomial

Normal Random Variables (4.3)

- A random variable is said to follow a **normal distribution** with **parameters** μ and σ if its pdf is:

Normal(μ, σ) Pdf:

$$f(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}} \quad \text{for } -\infty < x < \infty$$

- We write

$$X \sim \mathbf{N}(\mu, \sigma)$$

when X follows a normal distribution.

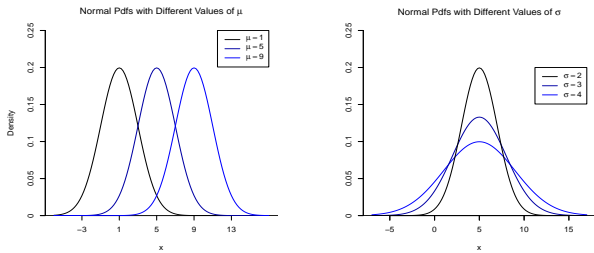
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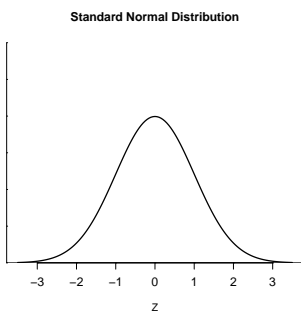


Normal(μ, σ) Mean and Variance: If $X \sim N(\mu, \sigma)$, then

$$E(X) = \mu$$

$$V(X) = \sigma^2$$

- The $N(0, 1)$ distribution ($\mu = 0$ and $\sigma = 1$) is called the **standard normal** distribution.



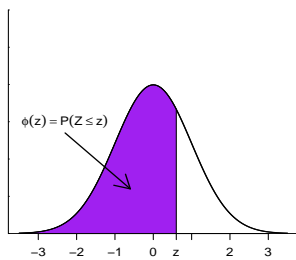
- We use Z to denote a **standard normal** random variable.
- The **pdf** of Z is

$$f(z) = \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}} \quad \text{for } -\infty < z < \infty$$

- The **cdf** of Z is denoted by $\phi(z)$. Thus

$$\phi(z) = P(Z \leq z),$$

Standard Normal Distribution



- To find probabilities such as $P(a \leq Z \leq b)$, we can't integrate the pdf,

$$\int_a^b f(z) dz = \int_a^b \frac{1}{\sqrt{2\pi}} e^{-\frac{z^2}{2}} dz$$

- Instead, the probabilities are obtained from values of $\phi(z) = P(Z \leq z)$ given in a **standard normal table**.

Example

Suppose

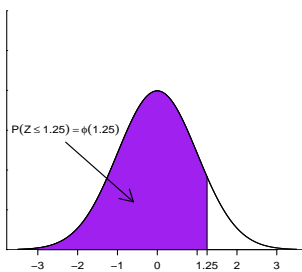
$$Z \sim N(0, 1)$$

From the **standard normal table**,

$$P(Z \leq 1.25) = \phi(1.25) = \mathbf{0.8944}$$

(1.2 row, 0.05 column of the table).

Standard Normal Distribution



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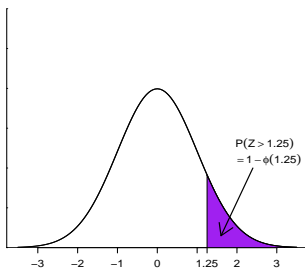
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Example

Also from the **standard normal table**,

$$P(Z > 1.25) = 1 - \phi(1.25) = 1 - 0.8944 = \mathbf{0.1056}.$$

Standard Normal Distribution

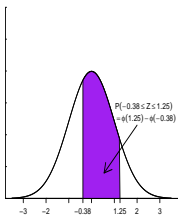


Example

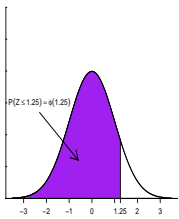
Also from the **standard normal table**,

$$\begin{aligned} P(-0.38 \leq Z \leq 1.25) &= \phi(1.25) - \phi(-0.38) \\ &= 0.8944 - 0.3520 \\ &= \mathbf{0.5424}. \end{aligned}$$

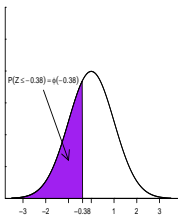
Standard Normal Distribution



Standard Normal Distribution



Standard Normal Distribution



- We just saw how to find probabilities using a $N(0, 1)$ distribution.
- To find probabilities using normal distributions with **other** values of μ and σ (besides 0 and 1), we use the following proposition.

Proposition

If $X \sim N(\mu, \sigma)$ and we define a new random variable Z by

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

then

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

Thus

$$1. P(X \leq a) = P\left(\frac{X - \mu}{\sigma} \leq \frac{a - \mu}{\sigma}\right) = P\left(Z \leq \frac{a - \mu}{\sigma}\right) = \Phi\left(\frac{a - \mu}{\sigma}\right)$$

$$1. P(X > b) = P\left(Z > \frac{b - \mu}{\sigma}\right) = 1 - \Phi\left(\frac{b - \mu}{\sigma}\right)$$

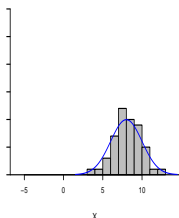
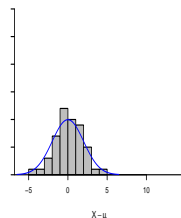
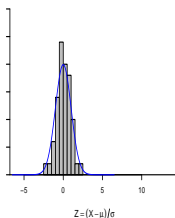
$$2. P(a \leq X \leq b) = P\left(\frac{a - \mu}{\sigma} \leq Z \leq \frac{b - \mu}{\sigma}\right) = \Phi\left(\frac{b - \mu}{\sigma}\right) - \Phi\left(\frac{a - \mu}{\sigma}\right)$$

- The variable

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

is called a **standardized** version of X , or **z-score**.

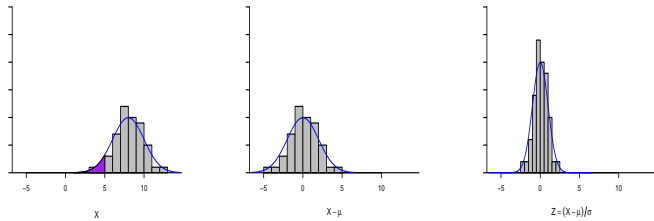
It's measured in **standard units** (standard deviations away from the mean).

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

Histogram of X

Histogram of $X-\mu$

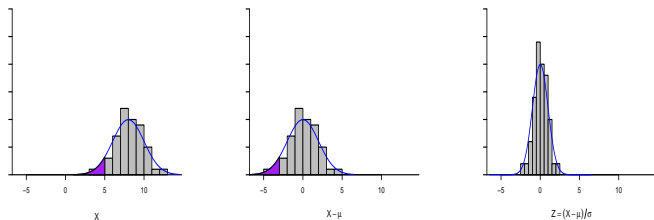
Histogram of $Z = \frac{X-\mu}{\sigma}$



Histogram of X

Histogram of $X-\mu$

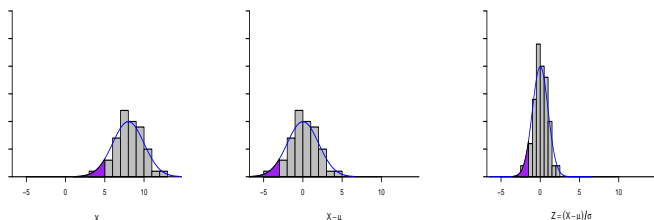
Histogram of $Z = \frac{X-\mu}{\sigma}$



Histogram of X

Histogram of $X-\mu$

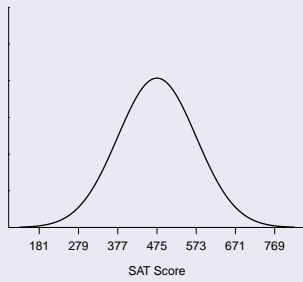
Histogram of $Z = \frac{X-\mu}{\sigma}$



Example

Scores on the verbal Scholastic Aptitude Test (SAT) follow a **normal distribution** with **mean 475** and **standard deviation 98**.

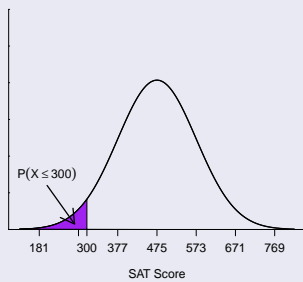
Normal Distribution of SAT Scores



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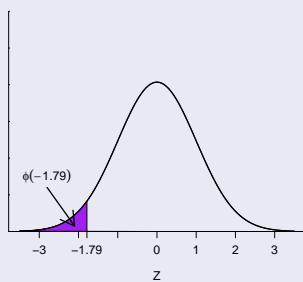
First we'll find $P(X \leq 300)$, the probability that an SAT score will be less than 300.

Normal Distribution of SAT Scores



Notes

Standard Normal Distribution



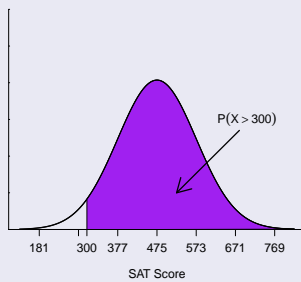
Notes

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$$\begin{aligned}
 P(X \leq 300) &= P\left(\frac{X-\mu}{\sigma} \leq \frac{300-\mu}{\sigma}\right) \\
 &= P\left(Z \leq \frac{300-475}{98}\right) \\
 &= P(Z \leq -1.79) \\
 &= \phi(-1.79) \\
 &= \mathbf{0.0367}.
 \end{aligned}$$

Now we'll find $P(X > 300)$, the probability that an SAT score will be **greater than 300**.

Normal Distribution of SAT Scores



$$\begin{aligned}
 P(X > 300) &= 1 - P(X \leq 300) \\
 &\vdots \\
 &= 1 - \phi(-1.79) \\
 &= 1 - 0.0367 \\
 &= \mathbf{0.9633}.
 \end{aligned}$$

Notes

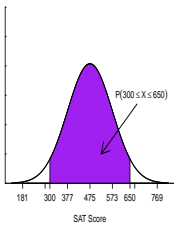
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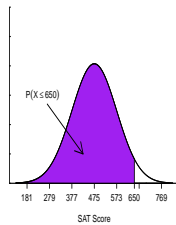
Notes

Lastly, we'll find $P(300 \leq X \leq 650)$, the probability that an SAT score will be **between 300 and 650**.

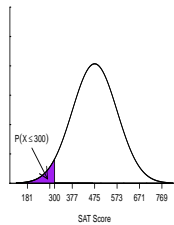
Normal Distribution of SAT Scores



Normal Distribution of SAT Scores



Normal Distribution of SAT Scores



$$\begin{aligned}
 P(300 \leq X \leq 650) &= P\left(\frac{300-\mu}{\sigma} \leq \frac{X-\mu}{\sigma} \leq \frac{650-\mu}{\sigma}\right) \\
 &= P\left(\frac{300-475}{98} \leq Z \leq \frac{650-475}{98}\right) \\
 &= P(-1.79 \leq Z \leq 1.79) \\
 &= \phi(1.79) - \phi(-1.79) \\
 &= 0.9633 - 0.0367 \\
 &= \mathbf{0.9266}.
 \end{aligned}$$

- A **standardized value** (or **z-score**) can be used to indicate an individual's standing **relative to others** in the population.

Example

Suppose you score **70** on your **Math** test, for which the **mean** is **65** and **standard deviation** is **5**.

Suppose also you score **80** on your **English** test, for which the **mean** is **75** and **standard deviation** is **7**.

On which test did you perform better relative to the rest of the class?

The **standardized** Math score is

$$Z = \frac{70 - 65}{5} = 1.0,$$

so it's **1.0** standard deviation above the mean.

The **standardized** English score is

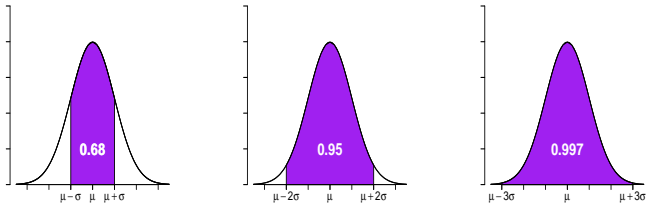
$$Z = \frac{80 - 75}{7} = 0.7,$$

so it's **0.7** of a standard deviation above the mean.

You did better on the **Math** test.

Empirical Rule (or 68-95-99.7 Rule): For any normal distribution,

1. Approximately **68%** of the distribution lies within **one σ of μ** .
2. Approximately **95%** of the distribution lies within **two σ 's of μ** .
3. Approximately **99.7%** of the distribution lies within **three σ 's of μ** .



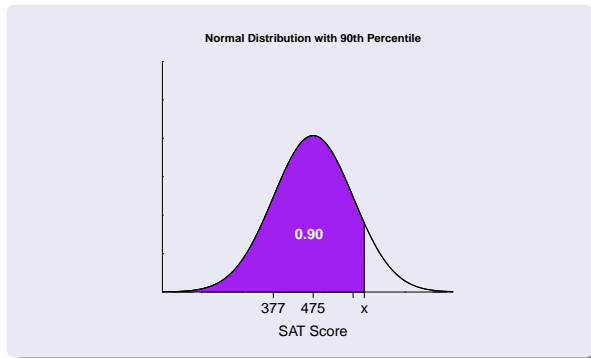
- The **(100p)th percentile** of a normal distribution is a value below which **(100p)%** of the distribution lies.
- For example, the **90th percentile** is the value below which **90%** of the distribution lies.

Example (Cont'd)

Recall that scores on the verbal SAT follow a **normal distribution** with **mean 475** and **standard deviation 98**.

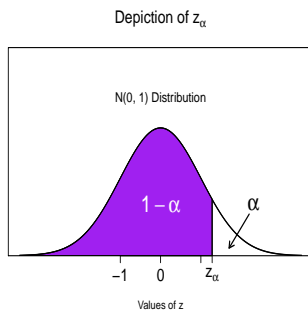
The **90th percentile** is the score below which **90%** of all scores lie.

It's marked x on the horizontal axis below.

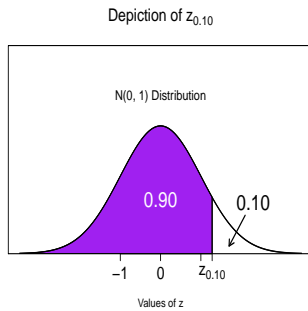


- We'll see how to find a **percentile** of a normal distribution with mean μ and standard deviation σ .
- First, though, we need to look at how to find a **percentile** of the **standard normal** distribution.

- We use z_α to denote the value that has area α to its **right** under the **$N(0, 1)$** curve.



- For example, $z_{0.10}$ has area **0.10** to its **right** under the $N(0, 1)$ curve:



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- z_α is the $100(1 - \alpha)$ th percentile of the $N(0, 1)$ distribution.
For example, $z_{0.10}$ is the **90th percentile** of the $N(0, 1)$ distribution.
- z_α is called a z **critical value**.

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Example

To find a z_α value, we search the body of the standard normal table for $1 - \alpha$, then get the corresponding z value (from the table margin). We find that

$$z_{0.10} = 1.28$$

$$z_{0.05} = 1.64$$

$$z_{0.025} = 1.96$$

$$z_{0.005} = 2.58$$

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The **100 p th percentile** of a $N(\mu, \sigma)$ distribution, x , is obtained by "unstandardizing" the 100 p th percentile z of the $N(0, 1)$ distribution:

$$x = \mu + z\sigma$$

The above expression was obtained by solving

$$z = \frac{x - \mu}{\sigma}$$

for x .

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Notes

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Example (Cont'd)

Recall that verbal SAT scores follow a **normal distribution** with **mean 475** and **standard deviation 98**.

We'll find the **90th percentile** of the distribution.

The 90th percentile of the **N(0, 1)** distribution is

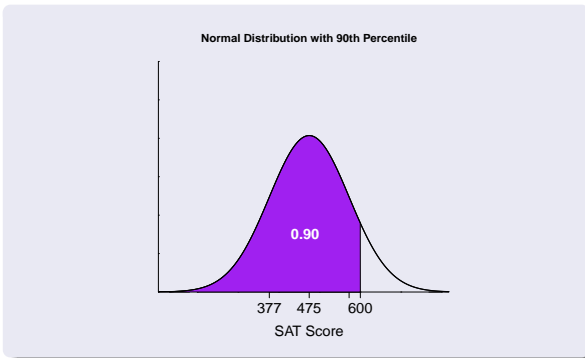
$$z = 1.28,$$

so the **90th percentile** of the distribution of **SAT scores** is

$$\begin{aligned} x &= \mu + z\sigma \\ &= 475 + 1.28(98) \\ &= \mathbf{600}. \end{aligned}$$

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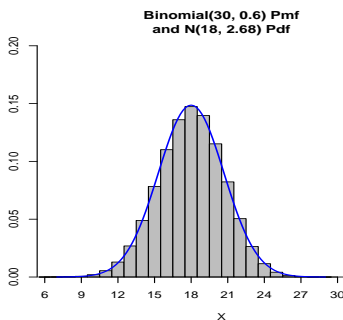
Notes



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- The **normal** distribution can be used to **approximate** a **binomial** distribution when the number of trials **n is large**.



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Proposition

Suppose

$$X \sim \text{binomial}(n, p),$$

so that $\mu_x = np$ and $\sigma_x = \sqrt{np(1-p)}$. Then if **n is large**,

$$X \sim \mathbf{N}\left(np, \sqrt{np(1-p)}\right) \quad \text{approximately.}$$

In particular, for each possible value x of X ,

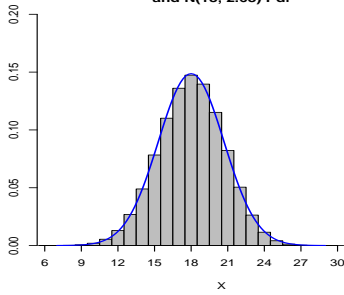
$$\begin{aligned} P(X \leq x) &\approx \left(\begin{array}{l} \text{Area under the normal curve} \\ \text{to the left of } x + 0.5 \end{array} \right) \\ &= \phi\left(\frac{x + 0.5 - np}{\sqrt{np(1-p)}}\right) \end{aligned}$$

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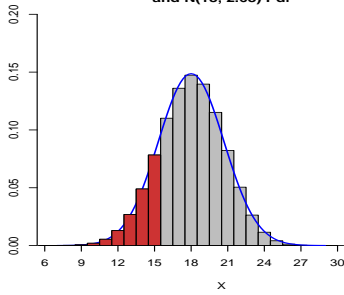
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- Adding **0.5** to x is referred to as a **continuity correction**.

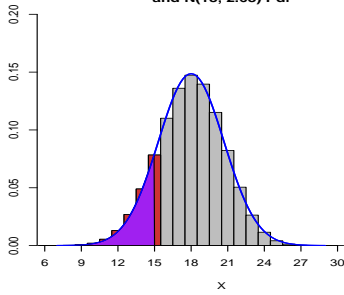
**Binomial(30, 0.6) Pmf
and N(18, 2.68) Pdf**

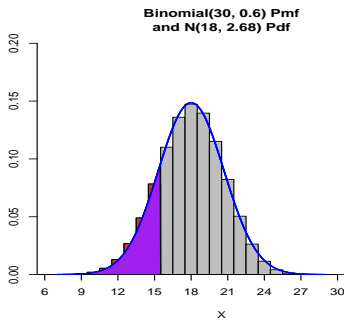


**Binomial(30, 0.6) Pmf
and N(18, 2.68) Pdf**



**Binomial(30, 0.6) Pmf
and N(18, 2.68) Pdf**



**Example**

A student is taking a true/false test with **100** questions. Suppose she has a probability $p = 3/4$ of getting each question right.

Let

X = The number of questions she gets right.

Then

$$X \sim \text{binomial}(100, 3/4)$$

We'll use the **normal distribution** to **approximate** the probability $P(X \leq 70)$ that she'll get **70 or fewer** right.

The **mean** and **standard deviation** of the distribution of X are

$$\mu_x = np = 100 \left(\frac{3}{4} \right) = 75$$

and

$$\sigma_x = \sqrt{np(1-p)} = \sqrt{100 \left(\frac{3}{4} \right) \left(\frac{1}{4} \right)} = 4.3$$

Because n is large, the **normal approximation to the binomial** gives

$$\begin{aligned} P(X \leq 70) &\approx \phi \left(\frac{70 + 0.5 - np}{\sqrt{np(1-p)}} \right) \\ &= \phi \left(\frac{70.5 - 75}{4.3} \right) \\ &= \phi(-1.05) \\ &= \mathbf{0.1469}. \end{aligned}$$

(Note that the exact binomial probability is 0.1495).

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- In practice, n is large enough for the **normal approximation to the binomial** to be valid as long as

$$np \geq 10 \quad \text{and} \quad n(1 - p) \geq 10.$$

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