

# STANDARD NORMAL DISTRIBUTION [DEVORE 9.1]

## • STANDARD NORMAL DISTRIBUTION (DEFINITION):

Notation	$Z \sim \text{Normal}(0, 1)$ or $Z \sim \text{StdNormal}$
Parameters	(None)
Support	$\text{Supp}(Z) = (-\infty, \infty)$
pdf	$f_Z(z) := \frac{1}{\sqrt{2\pi}} e^{-z^2/2}$
cdf	$\Phi(z) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^z e^{-\xi^2/2} d\xi$
Mean	$\mathbb{E}[Z] = 0$
Variance	$\mathbb{V}[Z] = 1$
Model(s)	(Used mainly for Statistical Inference)

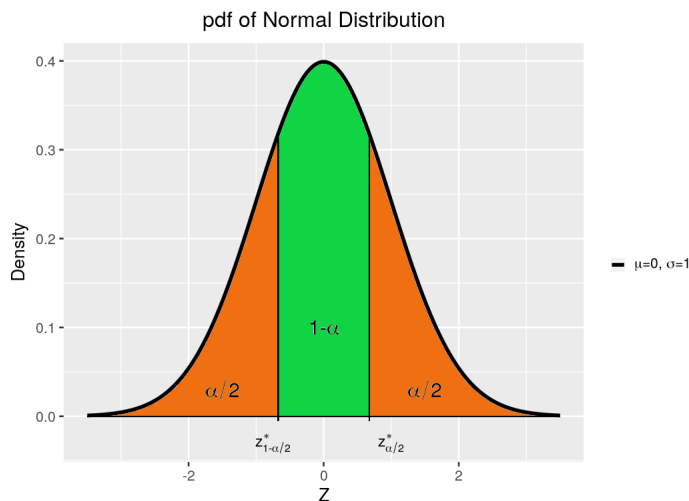
## • STANDARD NORMAL DISTRIBUTION (PROPERTIES):

- The standard normal pdf curve is unimodal.
- The standard normal pdf curve is bell-shaped.
- The standard normal pdf curve is symmetric.

## • z-CUTOFFS (AKA z CRITICAL VALUES):

- Value  $z_\alpha^*$  is a **z-cutoff (AKA z critical value)** of std normal distribution such that  $\mathbb{P}(Z > z_\alpha^*) = \alpha$
- NOTE: Do not confuse z-cutoff  $z_\alpha^*$  with z percentile  $z_\alpha$ :  $\mathbb{P}(Z \leq z_\alpha) = \alpha$
- Lower-tail z-cutoffs can be found from upper-tail z-cutoffs:  $z_{1-\alpha}^* = -z_\alpha^*$  (this follows since std normal is symmetric)

## • STANDARD NORMAL DISTRIBUTION (EXAMPLE PDF PLOT):



**STD NORMAL z-CUTOFFS,  $z_\alpha^*$**       $\mathbb{P}(Z > z_\alpha^*) = \alpha$ ,      $z_{1-\alpha}^* = -z_\alpha^*$

$\alpha$	<b>0.2</b>	<b>0.1</b>	<b>0.05</b>	<b>0.025</b>	<b>0.02</b>	<b>0.01</b>	<b>0.005</b>	<b>0.001</b>	<b>0.0005</b>
$z_\alpha^*$	0.842	1.282	1.645	1.960	2.054	2.326	2.576	3.090	3.291

# LARGE-SAMPLE $z$ -TESTS & $z$ -CI'S FOR $\mu_1 - \mu_2$ [DEVORE 9.1]

• **A STATISTIC RELATED TO THE STANDARD NORMAL DISTRIBUTION:**

Given any two populations with means  $\mu_1, \mu_2$  and unknown  $\sigma_1, \sigma_2$ .

Let  $\mathbf{X} := (X_1, \dots, X_{n_1})$  &  $\mathbf{Y} := (Y_1, \dots, Y_{n_2})$  be random samples from the 1<sup>st</sup> & 2<sup>nd</sup> populations, respectively.

Moreover, suppose random samples  $\mathbf{X}$  &  $\mathbf{Y}$  are independent of each other and their sizes are “large”, meaning  $n_1, n_2 > 40$ .

Then the following statistic has an approximate std normal distribution:  $\frac{(\bar{X} - \bar{Y}) - (\mu_1 - \mu_2)}{\sqrt{\frac{S_1^2}{n_1} + \frac{S_2^2}{n_2}}} \underset{\text{approx}}{\sim} \text{StdNormal}$

• **LARGE-SAMPLE  $z$ -TESTS FOR  $\mu_1 - \mu_2$ :**

Population:	Any Two Populations with unknown $\sigma_1, \sigma_2$
Realized Samples:	$\mathbf{x} := (x_1, x_2, \dots, x_{n_1})$ , mean $\bar{x}$ , std dev $s_1$ ( $n_1 > 40$ ) $\mathbf{y} := (y_1, y_2, \dots, y_{n_2})$ , mean $\bar{y}$ , std dev $s_2$ ( $n_2 > 40$ ) Samples $\mathbf{x}$ & $\mathbf{y}$ are independent of each other

Test Statistic Value: $W(\mathbf{x}, \mathbf{y}; \delta_0)$	$z = \frac{(\bar{x} - \bar{y}) - \delta_0}{\sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}}$
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HYPOTHESIS TEST:	REJECTION REGION AT LVL $\alpha$ :	P-VALUE DETERMINATION:
$H_0 : \mu_1 - \mu_2 = \delta_0$ vs. $H_A : \mu_1 - \mu_2 > \delta_0$	$z \geq z_\alpha^*$	P-value = $1 - \Phi(z)$
$H_0 : \mu_1 - \mu_2 = \delta_0$ vs. $H_A : \mu_1 - \mu_2 < \delta_0$	$z \leq z_{1-\alpha}^*$	P-value = $\Phi(z)$
$H_0 : \mu_1 - \mu_2 = \delta_0$ vs. $H_A : \mu_1 - \mu_2 \neq \delta_0$	$z \leq z_{1-\alpha/2}^*$ or $z \geq z_{\alpha/2}^*$	P-value = $2 \cdot [1 - \Phi( z )]$

**DECISION RULE:** If P-value  $\leq \alpha$  then reject  $H_0$  in favor of  $H_A$   
If P-value  $> \alpha$  then accept  $H_0$  (i.e. fail to reject  $H_0$ )

• **LARGE-SAMPLE  $z$ -CI'S FOR  $\mu_1 - \mu_2$ :**

Given any two populations with means  $\mu_1, \mu_2$  and unknown  $\sigma_1, \sigma_2$ .

Let  $\mathbf{x} := (x_1, x_2, \dots, x_{n_1})$  be a sample taken from the 1<sup>st</sup> population with  $n_1 > 40$ .

Let  $\mathbf{y} := (y_1, y_2, \dots, y_{n_2})$  be a sample taken from the 2<sup>nd</sup> population with  $n_2 > 40$ .

Moreover, suppose samples  $\mathbf{x}$  &  $\mathbf{y}$  are independent of each other.

Then the  $100(1 - \alpha)\%$  **large-sample  $z$ -CI** for  $\mu_1 - \mu_2$  is

$$\left( (\bar{x} - \bar{y}) - z_{\alpha/2}^* \cdot \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}, (\bar{x} - \bar{y}) + z_{\alpha/2}^* \cdot \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}} \right)$$

— OR WRITTEN MORE COMPACTLY —

$$(\bar{x} - \bar{y}) \pm z_{\alpha/2}^* \cdot \sqrt{\frac{s_1^2}{n_1} + \frac{s_2^2}{n_2}}$$

**EX 9.1.1:** A manufacturing plant produces a certain type of brake pad used in big rig trucks.

A sample of 55 day-shift workers showed that the mean # of brake pads produced was 310 with a std dev of 20.

A sample of 60 night-shift workers showed that the mean # of brake pads produced was 325 with a std dev of 26.

Assume that the two samples are independent.

Do these samples suggest that the average # brake pads produced by the two shifts differ?

(Use significance level  $\alpha = 0.05$ )

(a) State the appropriate null hypothesis  $H_0$  & alternative hypothesis  $H_A$ .

(b) Compute the appropriate  $z$ -test statistic value for this hypothesis test.

(c) Identify the two appropriate  $z$ -cutoffs and lookup/compute their values.

(d) Compute the appropriate P-value using software.

(e) Using the computed  $z$ -cutoffs, render the appropriate decision.

(f) Using the computed P-value, render the appropriate decision.

(g) Construct the approximate 95%  $z$ -CI for  $\mu_1 - \mu_2$ .