

STAT 113

Standardized Statistics

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Outline

Standard Normal

CIs from a Standard Normal

P-values Using a Standard Normal

Goals

Confidence Intervals

If we can replace the bootstrap distribution with a Normal model, we can construct a confidence interval.

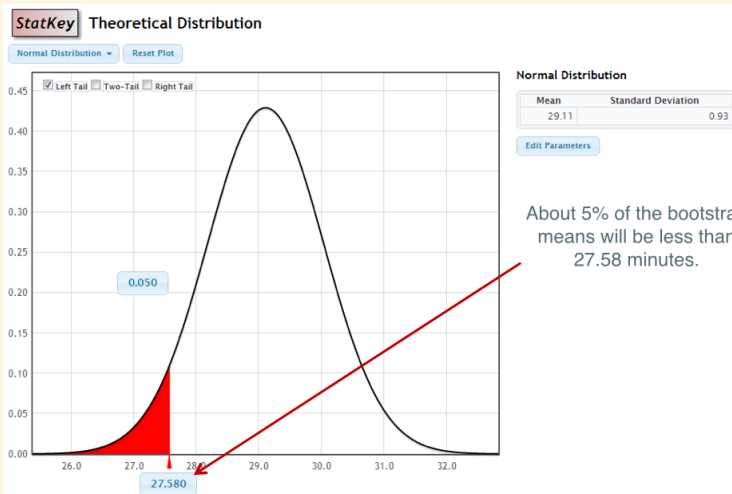
P-values

If we can replace a randomization distribution with a Normal model, we can compute *P*-values.

Quantiles of a Normal Curve

Suppose that the bootstrap distribution of means for samples of size 500 Atlanta commute times is $\mathcal{N}(29.11, 0.93)$. Find an endpoint (percentile) so that just 5% of the bootstrap means are smaller.

StatKey...

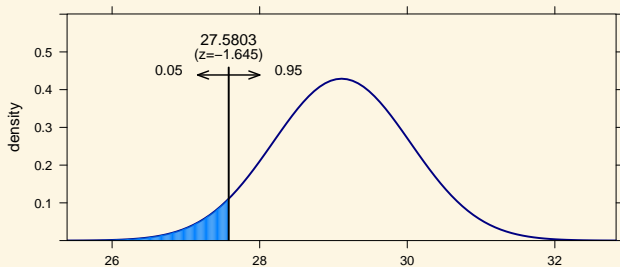


And in R ...

```
xqnorm(0.05, mean = 29.11, sd = 0.93)
```

```
## P(X <= 27.5802861269351) = 0.05
```

```
## P(X > 27.5802861269351) = 0.95
```



```
## [1] 27.58029
```

P-values Using a Normal

The mean commute time in the sample of 500 Atlanta commuters is 29.11 minutes. Is there evidence that the mean commute time for *all* Atlanta commuters is less than 30 minutes?

$$H_0 : \mu = 30$$

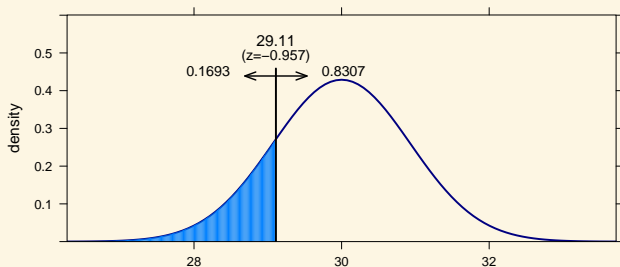
$$H_1 : \mu \neq 30$$

Suppose we can model the randomization distribution using a Normal with a standard error of 0.93. What should the mean be? Find the *P*-value.

In R ...

```
xpnorm(29.11, mean = 30, sd = 0.93)

##
## If  $X \sim N(30, 0.93)$ , then
##
##  $P(X \leq 29.11) = P(Z \leq -0.9569892) = 0.1692863$ 
##  $P(X > 29.11) = P(Z > -0.9569892) = 0.8307137$ 
```



```
## [1] 0.1692863
```


Outline

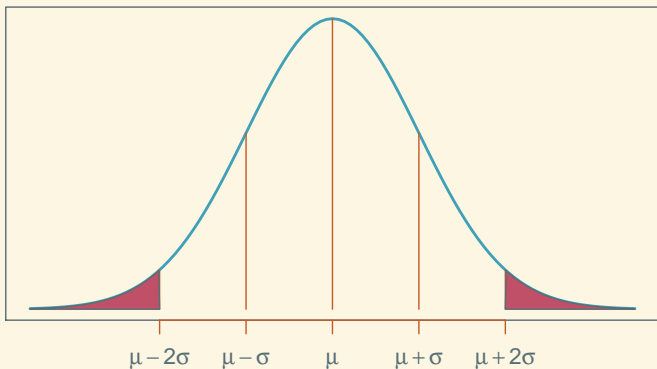
Standard Normal

CIs from a Standard Normal

P-values Using a Standard Normal

Quantiles of Normal Curves

The shape of a Normal is the same for all μ and σ . The mean is always at the peak; the “inflection points” are always $\mu + \sigma$ and $\mu - \sigma$, and 95% of the area is always between $\mu - 2\sigma$ and $\mu + 2\sigma$.



So, for proportions and quantiles, only “standard distances from the mean” (z -scores) matter!

What is a z -score?

The z -score for a point tells you how many standard deviations above the mean it is (negative = below)

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

If we relabel the x -axis of our density curve with a z -axis, we get what's called a **Standard Normal** distribution.

Normal and Standard Normal

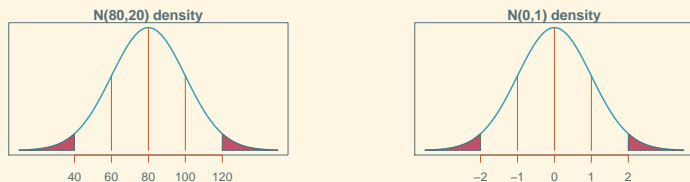


Figure: Left: Normal density with mean 80 and standard deviation 20.
Right: Standard Normal (mean 0, standard deviation 1).

Example: Gestation Time

Dear Abby: You wrote that a woman is pregnant for 266 days. Who said so? I carried my baby for ten months and five days, and there is no doubt about it because I know the exact date my baby was conceived. My husband is in the Navy and it couldn't have possibly been conceived any other time because I saw him only once for an hour, and I didn't see him again until the day before the baby was born.

I don't drink or run around, and there is no way the baby isn't his, so please print a retraction about the 266-day carrying time because otherwise I'm in a lot of trouble.

San Diego Reader

Dear San Diego Reader: Some babies come early, some come late; yours came late.

Abby

Example: Gestation Time

Human gestation times in days are distributed approximately $\mathcal{N}(266, 16)$. The reader was pregnant for 305 days.

- What is that as a z -score?
- Use the raw score to find the reader's percentile.
- Use the z -score to find the reader's percentile.

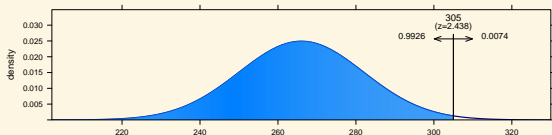
Solutions: Gestation Time

Human gestation times in days are distributed approximately $\mathcal{N}(266, 16)$. The reader was pregnant for 305 days.

$$z = \frac{X - \mu}{\sigma} = \frac{305 - 266}{16} = 2.4375$$

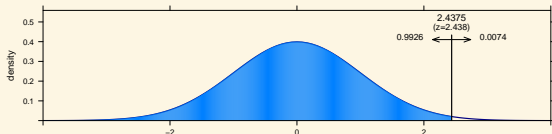
Solutions: Gestation Time

```
### Using the raw score, the percentile is given by xpnorm:  
xpnorm(305, mean = 266, sd = 16, lower.tail = TRUE, verbose = FALSE)
```



```
## [1] 0.9926054
```

```
### When we use the z score, we locate it in the standard normal:  
xpnorm(2.4375, mean = 0, sd = 1, lower.tail = TRUE, verbose = FALSE)
```



Outline

Standard Normal

CIs from a Standard Normal

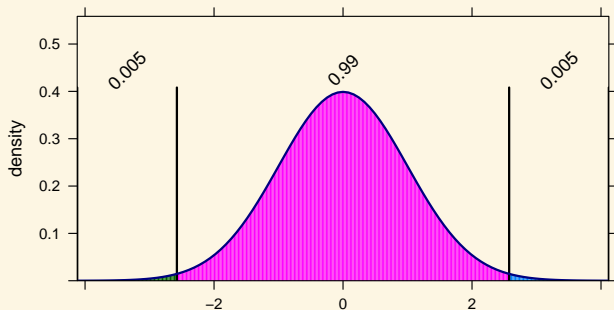
P-values Using a Standard Normal

Confidence Intervals from a Standard Normal

- We already know that Sample Statistic ± 2 SE yields an (approximately) 95% CI. What are the z -scores associated with these endpoints in the context of the bootstrap distribution?
- When the bootstrap distribution is Normal, the z -scores for a given confidence level are always the same.
 - 95%: $z \approx \pm 2$
 - 99%: ?
 - 90%: ?
- How can we find these using a standard normal?

Confidence Intervals from a Standard Normal

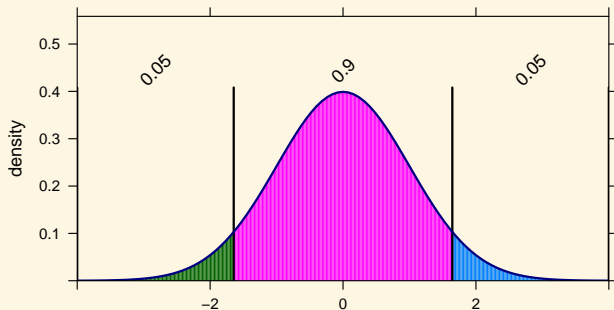
```
### Find the 0.005 and 0.995 quantiles of the standard Normal.  
### These are the z-scores of the 99% confidence interval  
### (within the bootstrap distribution)  
xqnorm(c(0.005, 0.995), mean = 0, sd = 1, verbose = FALSE)
```



```
## [1] -2.575829 2.575829
```

Confidence Intervals from a Standard Normal

```
### Find the 0.05 and 0.95 quantiles of the standard Normal.  
### These are the z-scores of the 90% confidence interval  
### (within the bootstrap distribution)  
xqnorm(c(0.05, 0.95), mean = 0, sd = 1, lower.tail = TRUE, verbose = FALSE)
```



```
## [1] -1.644854 1.644854
```

Z-score conversion

The relationship between the original scale and standardized scale is

$$Z = \frac{\text{Original} - \text{Distribution Mean}}{\text{Standard Deviation}}$$

Converting back to the original scale

If we find the z -scores of the CI endpoints, we can convert them to a confidence interval on the original scale.

$$\text{Endpoint(Original)} = \text{Distribution Mean} + Z \cdot \text{Standard Deviation}$$

Demo

Converting back to the original scale

If we find the z -scores of the CI endpoints, we can convert them to a confidence interval on the original scale.

$$\text{Endpoint(Original)} = \text{Distribution Mean} + Z \cdot \text{Standard Deviation}$$

CI Summary

To compute a confidence interval when the bootstrap distribution can be replaced by a Normal, use

$$\text{Endpoint} = \text{observed statistic} \pm Z^* \cdot \text{Bootstrap SE}$$

where Z^* is the Z -score of the endpoint appropriate for the confidence level, computed from a standard normal ($\mathcal{N}(0, 1)$).

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P-values from a Standard Normal

Computing P -values when the randomization distribution is Normal is the reverse process:

1. Convert the observed statistic to a z -score within the randomization distribution (i.e., using its mean and standard deviation).

$$Z_{observed} = \frac{\text{observed statistic} - \text{null parameter}}{\text{randomization SD}}$$

2. Find the relevant area beyond $Z_{observed}$ using a Standard Normal

Example: Sleep and Caffeine

Is mean number of words recalled different after sleep vs. caffeine?

$$H_0 : \mu_{\text{sleep}} - \mu_{\text{caffeine}} = 0$$

$$H_1 : \mu_{\text{sleep}} - \mu_{\text{caffeine}} \neq 0$$

Sample statistic: $\bar{x}_{\text{sleep}} - \bar{x}_{\text{caffeine}}$

P-value: Demo