

t -Test in R

MD MAHFUJUL KARIM SHEIKH

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Contents

Required Packages	1
<code>install.packages("reshape2")</code>	1
<code>library(reshape2)</code>	1
One sample t-test	2
Manual Calculation	2
using function————	2
Using Function	4
Two-samples t-test	7
Paired samples t-test	9

Required Packages

`install.packages("reshape2")`

`library(reshape2)`

```
library(ggplot2)
library(reshape2)
library(car)
library(ggpubr)
```

One sample t-test

Manual Calculation

```
set.seed(123)
sample_data <- rnorm(30, mean = 50, sd = 10)
mean(sample_data)
```

```
## [1] 49.52896
```

```
sd(sample_data)
```

```
## [1] 9.810307
```

```
#mu_0=55
SE <- sd(sample_data)/sqrt(length(sample_data))
t_cal <- (mean(sample_data) - 55)/SE
print(t_cal)
```

```
## [1] -3.054553
```

```
t_crit <- qt(0.05/2, df = length(sample_data)-1, lower.tail=TRUE)
t_crit
```

```
## [1] -2.04523
```

```
t_cal <- t_crit
```

using function—————

```
t.test(sample_data, mu=55)
```

```
##
## One Sample t-test
##
## data: sample_data
## t = -3.0546, df = 29, p-value = 0.004797
## alternative hypothesis: true mean is not equal to 55
## 95 percent confidence interval:
## 45.86573 53.19219
## sample estimates:
## mean of x
## 49.52896
```

```
t.test(sample_data, mu=55, conf.level=0.99)
```

```
##  
## One Sample t-test  
##  
## data: sample_data  
## t = -3.0546, df = 29, p-value = 0.004797  
## alternative hypothesis: true mean is not equal to 55  
## 99 percent confidence interval:  
## 44.59198 54.46595  
## sample estimates:  
## mean of x  
## 49.52896
```

```
t.test(sample_data, mu=55, conf.level=0.99)
```

```
##  
## One Sample t-test  
##  
## data: sample_data  
## t = -3.0546, df = 29, p-value = 0.004797  
## alternative hypothesis: true mean is not equal to 55  
## 99 percent confidence interval:  
## 44.59198 54.46595  
## sample estimates:  
## mean of x  
## 49.52896
```

Hypothesis:

$H_0 : \mu = \mu_0$

$H_1 : \mu \neq \mu_0$

```
# mu_0 = 55  
SE <- sd(sample_data)/sqrt(length(sample_data))  
t_cal <- (mean(sample_data) - 55)/SE  
print(t_cal)
```

```
## [1] -3.054553
```

```
t_crit <- qt(0.05/2, df = length(sample_data)-1, lower.tail = TRUE)  
t_crit
```

```
## [1] -2.04523
```

```
t_cal <= t_crit # decision: reject null
```

```
## [1] TRUE
```

```
pt(t_cal, df = 29, lower.tail = TRUE) + pt(-t_cal, df = 29, lower.tail = FALSE)
```

```
## [1] 0.0047971
```

```
mean(sample_data)+c(-1, 1)* abs(t_crit)*SE
```

```
## [1] 45.86573 53.19219
```

Calculating confidence interval:

```
mean(sample_data) + c(-1, 1) * abs(t_crit) * SE
```

```
## [1] 45.86573 53.19219
```

Using Function

Generate sample data:

```
set.seed(123)
sample_data <- rnorm(30, mean = 50, sd = 10)
```

Testing normality using Shapiro-Wilk test:

```
shapiro.test(sample_data)
```

```
##
##  Shapiro-Wilk normality test
##
## data:  sample_data
## W = 0.97894, p-value = 0.7966
```

H_0 : Data follows normal distribution.

H_1 : Data does not follow normal distribution.

Since the p-value is greater than the level of significance ($\alpha = 0.05$), we do not have enough statistical evidence to reject the null hypothesis.

One-sample t-test:

Hypothesis:

$H_0 : \mu = 50$

$H_1 : \mu \neq 50$

```
t.test(sample_data, mu = 50)
```

```
##
##  One Sample t-test
##
## data:  sample_data
```

```
## t = -0.26299, df = 29, p-value = 0.7944
## alternative hypothesis: true mean is not equal to 50
## 95 percent confidence interval:
##  45.86573 53.19219
## sample estimates:
## mean of x
##  49.52896
```

Using function, perform the two tailed t-test:

```
t.test(sample_data, mu = 53, conf.level = 0.95)
```

```
##
## One Sample t-test
##
## data: sample_data
## t = -1.9379, df = 29, p-value = 0.06242
## alternative hypothesis: true mean is not equal to 53
## 95 percent confidence interval:
##  45.86573 53.19219
## sample estimates:
## mean of x
##  49.52896
```

If $p\text{-value} < 0.05$, then reject null. Decision is not rejected.

```
t.test(sample_data, mu = 53, conf.level = 0.99)
```

```
##
## One Sample t-test
##
## data: sample_data
## t = -1.9379, df = 29, p-value = 0.06242
## alternative hypothesis: true mean is not equal to 53
## 99 percent confidence interval:
##  44.59198 54.46595
## sample estimates:
## mean of x
##  49.52896
```

If $p\text{-value} < 0.01$, then reject null. Decision is not rejected.

```
t.test(sample_data, mu = 53, conf.level = 0.90)
```

```
##
## One Sample t-test
##
## data: sample_data
## t = -1.9379, df = 29, p-value = 0.06242
## alternative hypothesis: true mean is not equal to 53
## 90 percent confidence interval:
```

```
## 46.48564 52.57228
## sample estimates:
## mean of x
## 49.52896
```

If p-value < 0.10, then reject null. Decision is rejected.

Hypothesis:

$H_0 : \mu \leq 40$

$H_1 : \mu > 40$

```
t.test(sample_data, mu = 40, alternative = "greater")
```

```
##
## One Sample t-test
##
## data: sample_data
## t = 5.3201, df = 29, p-value = 5.21e-06
## alternative hypothesis: true mean is greater than 40
## 95 percent confidence interval:
## 46.48564 Inf
## sample estimates:
## mean of x
## 49.52896
```

Hypothesis:

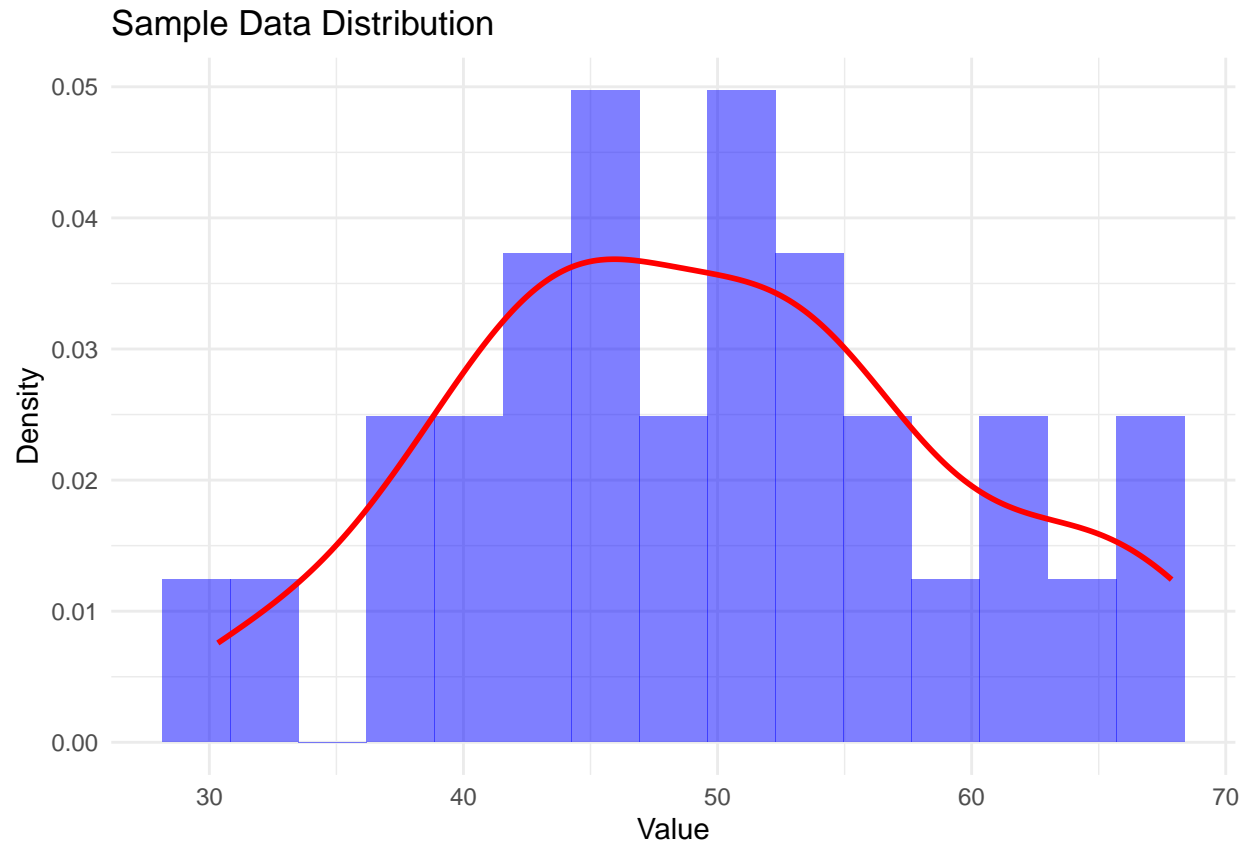
$H_0 : \mu \geq 58$

$H_1 : \mu < 58$

```
t.test(sample_data, mu = 58, alternative = "less")
```

```
##
## One Sample t-test
##
## data: sample_data
## t = -4.7295, df = 29, p-value = 2.689e-05
## alternative hypothesis: true mean is less than 58
## 95 percent confidence interval:
## -Inf 52.57228
## sample estimates:
## mean of x
## 49.52896
```

```
ggplot(data.frame(Value = sample_data), aes(x = Value)) +
  geom_histogram(aes(y = after_stat(density)), bins = 15, fill = "blue", alpha = 0.5) +
  geom_density(color = "red", linewidth = 1) +
  labs(title = "Sample Data Distribution", x = "Value", y = "Density") +
  theme_minimal()
```



Two-samples t-test

Use built-in dataset: mtcars (comparing mpg for automatic vs manual cars):

```
data(mtcars)
```

Split into two groups based on transmission type”:

```
auto_mpg <- mtcars$mpg[mtcars$am == 0] # Automatic  
manual_mpg <- mtcars$mpg[mtcars$am == 1] # Manual
```

H_0 : Automatic cars and manual cars have equal average mpg.

H_1 : Automatic cars and manual cars have unequal average mpg.

```
mean(auto_mpg)
```

```
## [1] 17.14737
```

```
mean(manual_mpg)
```

```
## [1] 24.39231
```

```
var(auto_mpg)
```

```
## [1] 14.6993
```

```
var(manual_mpg)
```

```
## [1] 38.02577
```

Normality test for both groups:

```
shapiro.test(auto_mpg)
```

```
##  
## Shapiro-Wilk normality test  
##  
## data: auto_mpg  
## W = 0.97677, p-value = 0.8987
```

```
shapiro.test(manual_mpg)
```

```
##  
## Shapiro-Wilk normality test  
##  
## data: manual_mpg  
## W = 0.9458, p-value = 0.5363
```

Check variance homogeneity (Levene's test):

```
leveneTest(mpg ~ factor(am), data = mtcars, center = "mean")
```

```
## Levene's Test for Homogeneity of Variance (center = "mean")  
##      Df F value Pr(>F)  
## group 1    5.921 0.02113 *  
##      30  
## ---  
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

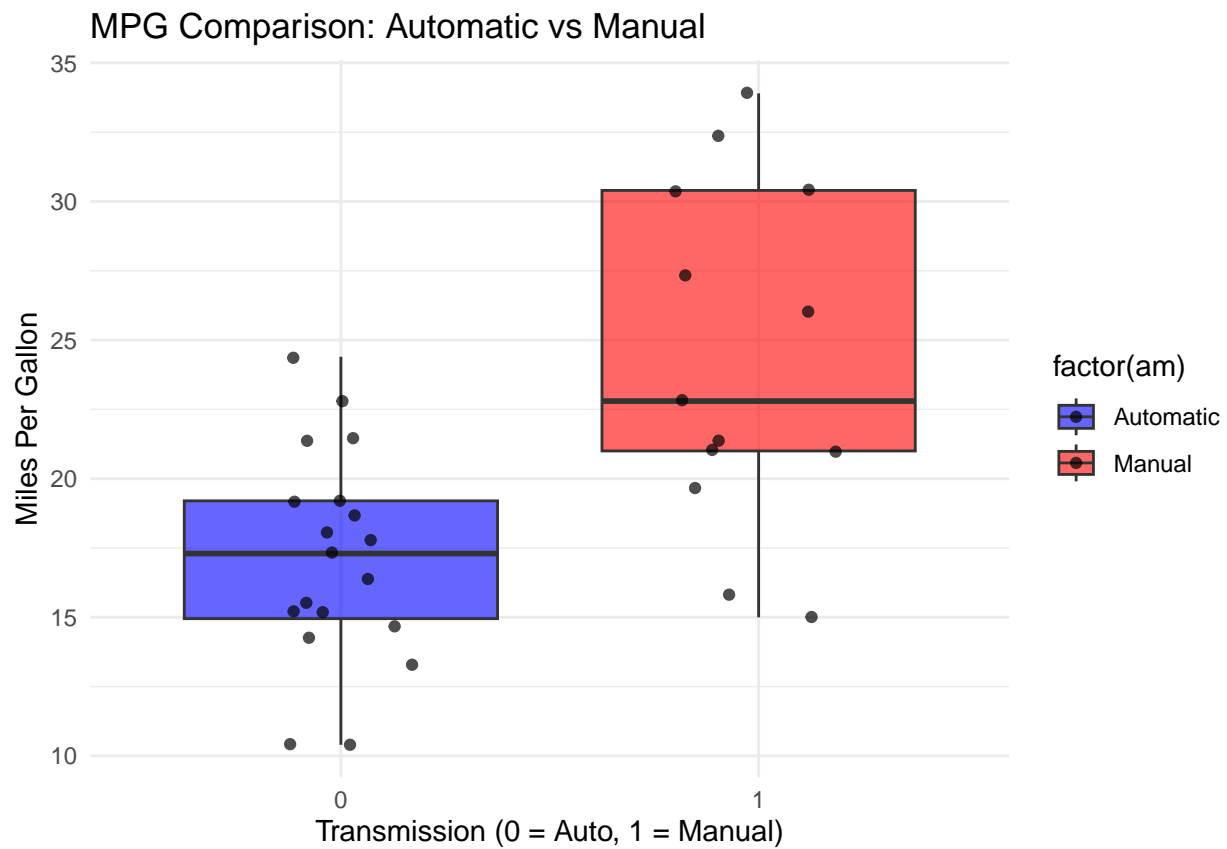
Perform two-sample t-test:

```
t.test(auto_mpg, manual_mpg, var.equal = FALSE)
```

```
##  
## Welch Two Sample t-test  
##  
## data: auto_mpg and manual_mpg  
## t = -3.7671, df = 18.332, p-value = 0.001374  
## alternative hypothesis: true difference in means is not equal to 0  
## 95 percent confidence interval:  
## -11.280194 -3.209684  
## sample estimates:  
## mean of x mean of y  
## 17.14737 24.39231
```



```
# Visualize the data
ggplot(mtcars, aes(x = factor(am), y = mpg, fill = factor(am))) +
  geom_boxplot(alpha = 0.6) +
  geom_jitter(width = 0.2, alpha = 0.7) +
  labs(title = "MPG Comparison: Automatic vs Manual",
       x = "Transmission (0 = Auto, 1 = Manual)",
       y = "Miles Per Gallon") +
  scale_fill_manual(values = c("blue", "red"),
                   labels = c("Automatic", "Manual")) +
  theme_minimal()
```



Paired samples t-test

Generate 30 observations:

```
set.seed(123)
before <- round(rnorm(30, mean = 300, sd = 10), 0)
summary(before)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##  280.0   293.2   299.5   299.6   305.0   318.0
```

```
# after a course
after <- before + round(rnorm(30, mean = 5, sd = 5), 0) # Simulating a increase
summary(after)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##    283.0   298.2   305.0   305.5   312.2   325.0
```

```
df <- data.frame(
  ID = 1:30,
  Before = before,
  After = after
)
```

```
mean(after) - mean(before)
```

```
## [1] 5.933333
```

H_0 : Before and after the course the true GRE average score of the students stays the same.

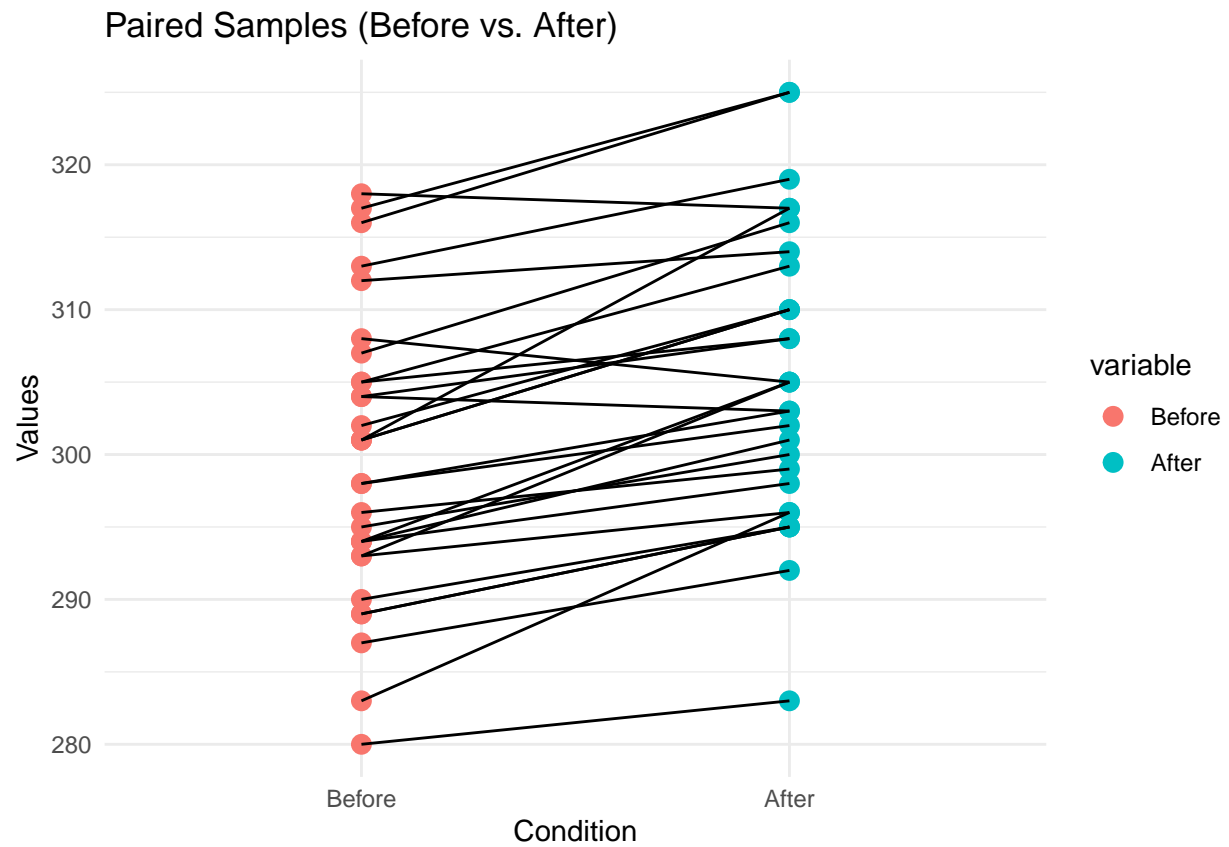
H_1 : Before and after the course the true GRE average score of the students does not remain the same.

Perform Paired t-test

```
t.test(x = after, y = before, paired = TRUE, alternative = "two.sided")
```

```
##
## Paired t-test
##
## data: after and before
## t = 7.7811, df = 29, p-value = 1.398e-08
## alternative hypothesis: true mean difference is not equal to 0
## 95 percent confidence interval:
##  4.373779 7.492888
## sample estimates:
## mean difference
##      5.933333
```

```
# Visualize the differences
df <- data.frame(
  ID = 1:30,
  Before = before,
  After = after
)
df_long <- melt(df, id.vars = "ID")
ggplot(df_long, aes(x = variable, y = value, group = ID)) +
  geom_point(aes(color = variable), size = 3) +
  geom_line() +
  labs(title = "Paired Samples (Before vs. After)",
       y = "Values", x = "Condition") +
  theme_minimal()
```



```
ggpaired(df_long,  
  x = "variable",  
  y = "value",  
  color = "variable",  
  line.color = "gray",  
  line.size = 0.4,  
  palette = "jco") +  
stat_compare_means(paired = TRUE, method = "t.test")
```

