


```
pnorm(z)
```

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

which is greater than our $\alpha = 0.05$. We cannot disprove H_0 .

Using the R function:

```
prop.test(sum(B_GPA), length(B_GPA), p = p0, alternative = c("less"), correct = FALSE)
```

```
##
## 1-sample proportions test without continuity correction
##
## data:  sum(B_GPA) out of length(B_GPA), null probability p0
## X-squared = 64.882, df = 1, p-value = 1
## alternative hypothesis: true p is less than 0.3
## 95 percent confidence interval:
##  0.0000000 0.7933579
## sample estimates:
##           p
## 0.7179487
```

Where $p = 1 > \alpha = 0.05$. We cannot disprove H_0 .

Problem 2

```
M_GPA <- GpaGender[which(GpaGender$Gender == 2), "GPA"]; M_GPA
```

```
## # A tibble: 47 x 1
##   GPA
##   <dbl>
## 1  7.94
## 2  8.29
## 3  4.64
## 4  7.47
## 5  7.58
## 6  7.65
## 7  2.41
## 8  8.83
## 9  7.17
## 10 4
## # i 37 more rows
```

```
F_GPA <- GpaGender[which(GpaGender$Gender == 1), "GPA"]; F_GPA
```

```
## # A tibble: 31 x 1
##   GPA
##   <dbl>
## 1  8.88
## 2  6
## 3  7.47
## 4  5.53
## 5  7.57
## 6  4.7
## 7  8.17
## 8  7.82
```

```
## 9 7.60
## 10 6.23
## # i 21 more rows
```

```
# Male summary:
summary(M_GPA)
```

```
##      GPA
## Min.   : 0.530
## 1st Qu.: 6.556
## Median : 7.825
## Mean    : 7.282
## 3rd Qu.: 9.000
## Max.    :10.760
```

```
# Female summary:
summary(F_GPA)
```

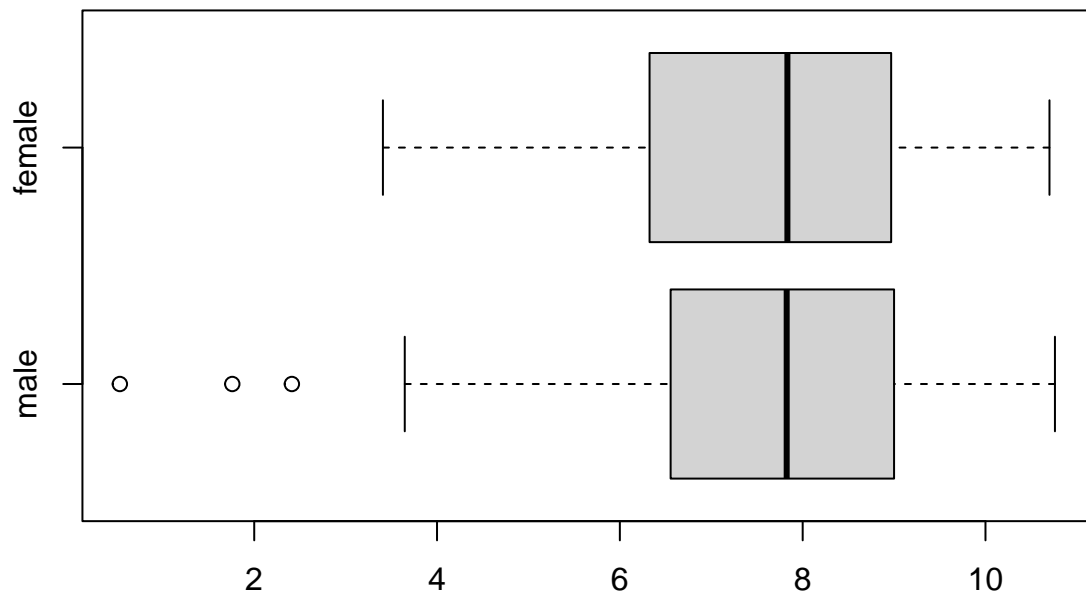
```
##      GPA
## Min.   : 3.408
## 1st Qu.: 6.325
## Median : 7.833
## Mean    : 7.697
## 3rd Qu.: 8.968
## Max.    :10.700
```

Problem 3

```
both_GPA <- c(M_GPA, F_GPA); both_GPA
```

```
## $GPA
## [1] 7.940 8.292 4.643 7.470 7.585 7.650 2.412 8.833 7.167 4.000
## [11] 7.643 1.760 9.648 10.580 9.429 8.000 9.585 8.175 8.000 9.500
## [21] 9.167 10.760 9.763 9.410 9.167 9.348 8.167 3.647 3.936 7.167
## [31] 7.647 0.530 6.173 7.295 8.353 5.062 8.175 8.235 7.588 7.647
## [41] 7.825 9.167 7.996 4.885 3.820 6.057 6.938
##
## $GPA
## [1] 8.882 6.000 7.470 5.528 7.571 4.700 8.167 7.822 7.598 6.231
## [11] 6.419 10.700 9.571 8.998 8.333 9.333 10.140 9.999 3.408 7.295
## [21] 8.938 7.882 5.237 7.333 8.714 7.833 7.998 5.936 9.000 9.500
## [31] 6.057
```

```
boxplot(both_GPA, names = c("male", "female"), horizontal = TRUE)
```



Problem 4

```
t.test(M_GPA, conf.level = 0.9)
```

```
##
## One Sample t-test
##
## data: M_GPA
## t = 21.527, df = 46, p-value < 2.2e-16
## alternative hypothesis: true mean is not equal to 0
## 90 percent confidence interval:
##  6.713815 7.849462
## sample estimates:
## mean of x
##  7.281638
```

```
t.test(F_GPA, conf.level = 0.9)
```

```
##
## One Sample t-test
##
## data: F_GPA
## t = 24.903, df = 30, p-value < 2.2e-16
## alternative hypothesis: true mean is not equal to 0
## 90 percent confidence interval:
##  7.171981 8.221116
```

```
## sample estimates:
## mean of x
## 7.696548
```

90% Confidence interval for males is [6.713815, 7.849462]; for females it is [7.171981, 8.221116].

Problem 5

```
M_GPA <- GpaGender[which(GpaGender$Gender == 2), "GPA"]$GPA; M_GPA
```

```
## [1] 7.940 8.292 4.643 7.470 7.585 7.650 2.412 8.833 7.167 4.000
## [11] 7.643 1.760 9.648 10.580 9.429 8.000 9.585 8.175 8.000 9.500
## [21] 9.167 10.760 9.763 9.410 9.167 9.348 8.167 3.647 3.936 7.167
## [31] 7.647 0.530 6.173 7.295 8.353 5.062 8.175 8.235 7.588 7.647
## [41] 7.825 9.167 7.996 4.885 3.820 6.057 6.938
```

```
F_GPA <- GpaGender[which(GpaGender$Gender == 1), "GPA"]$GPA; F_GPA
```

```
## [1] 8.882 6.000 7.470 5.528 7.571 4.700 8.167 7.822 7.598 6.231
## [11] 6.419 10.700 9.571 8.998 8.333 9.333 10.140 9.999 3.408 7.295
## [21] 8.938 7.882 5.237 7.333 8.714 7.833 7.998 5.936 9.000 9.500
## [31] 6.057
```

$H_0 : \sigma_M^2 = \sigma_F^2$ where $H_0 \sim F_{n_M-1, n_F-1} \equiv F_{46, 30}$

Let's check our sample variances:

```
SSM <- var(M_GPA); SSM
```

```
## [1] 5.377723
```

```
SSF <- var(F_GPA); SSF
```

```
## [1] 2.961198
```

Since $S_M^2 > S_F^2$, $H_a : \sigma_M^2 > \sigma_F^2$

Our observed $F = f =$

```
f <- SSM/SSF; f
```

```
## [1] 1.816063
```

and gives us p-value

```
1 - pf(f, length(M_GPA) - 1, length(F_GPA) - 1)
```

```
## [1] 0.04321679
```

which is less than $\alpha = 0.05$. We can reject H_0 and assume $\sigma_M^2 > \sigma_F^2$. Using the R function:

```
var.test(M_GPA, F_GPA, ratio = 1, alternative = c("greater"), conf.level = 1 - alpha)
```

```
##
```

```
## F test to compare two variances
```

```
##
```

```
## data: M_GPA and F_GPA
```

```
## F = 1.8161, num df = 46, denom df = 30, p-value = 0.04322
```

```
## alternative hypothesis: true ratio of variances is greater than 1
```

```
## 95 percent confidence interval:
```

```
## 1.025001 Inf
```

```
## sample estimates:
## ratio of variances
##          1.816063
```

Problem 6

We know $\sigma_M^2 \neq \sigma_F^2$ and we know neither value. So we will do a general two-sample T-test.

$$H_0 : \mu_M = \mu_F$$

$$H_a : \mu_M \neq \mu_F$$

We calculate our observed $T = t$ where

```
xbar <- mean(M_GPA); xbar
```

```
## [1] 7.281638
```

```
ybar <- mean(F_GPA); ybar
```

```
## [1] 7.696548
```

```
n1 <- length(M_GPA); n1
```

```
## [1] 47
```

```
n2 <- length(F_GPA); n2
```

```
## [1] 31
```

```
t <- (xbar - ybar)/sqrt(SSM/n1 + SSF/n2); t
```

```
## [1] -0.9055327
```

and get a p-value with

```
k <- (SSM/n1+SSF/n2)^2/((SSM/n1)^2/(n1-1)+(SSF/n2)^2/(n2-1))
k <- ceiling(k); k
```

```
## [1] 75
```

```
2 * pt(-abs(t), k)
```

```
## [1] 0.3680831
```

which is greater than 0.05, so we cannot reject H_0 , so $\mu_M = \mu_F$.

Using the R function:

```
t.test(M_GPA, F_GPA, alternative = c("two.sided"), paired = FALSE, var.equal = FALSE)
```

```
##
## Welch Two Sample t-test
##
## data: M_GPA and F_GPA
## t = -0.90553, df = 74.862, p-value = 0.3681
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -1.3277078 0.4978876
## sample estimates:
## mean of x mean of y
## 7.281638 7.696548
```