

Solution to Homework 4

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Warning: This note is only used as a reference solution for the homework, and the solution to each question is not unique. The solution may contain factual and/or typographic errors and comments and criticism are kindly welcomed.

Remark: In Problem 1 and 3, to compare "which test is more powerful" you'd better calculate the power of test using simulation (which is omitted in solution and you can try it on your own). Tips: You can estimate the power by repeating the tests under specific data setting and take the average reject rate as an estimation.

Problem 1 Let $X_1, \dots, X_n \stackrel{iid}{\sim} N(0, \sigma_X^2)$ and $Y_1, \dots, Y_n \stackrel{iid}{\sim} N(0, \sigma_Y^2)$. Use simulation to answer the following: 1) Suppose $n = 100$, $\sigma_X^2 = 1$ and $\sigma_Y^2 = 1.2$, use the Mood test and the Sukhatme test to test $H_0 : \sigma_X^2 = \sigma_Y^2$ at level $\alpha = 0.05$. 2) Can you think of a scenario where the Mood test is more powerful than the Sukhatme test? 3) Can you think of a scenario where the Sukhatme test is more powerful than the Mood test?

Solution:

```

1 mood.stats <- function(X, Y) {
2   X <- sort(X); Y <- sort(Y)
3   XY <- sort(c(X, Y))
4   rank.X <- match(X, XY)
5   return(sum((rank.X - length(XY) + 1) / 2) ^ 2)
6 }

```

```
7
8 moods.pvalue <- function(X,Y,nrep){
9   # Calculate the Mood test statistics
10  n <- length(X)
11  stats <- mood.stats(X, Y)
12  # Simulate under the null
13  mood.sim <- c()
14  for (i in 1:nrep) {
15    mood.sim[i] <- mood.stats(rnorm(n,0,1), rnorm(n,0,1))
16  }
17  return(2*min(mean(mood.sim < stats), mean(mood.sim > stats)))
18 }
19
20 Sukhatme.stats <- function(X,Y){
21   data.pos <- merge(X[which(X>0)],Y[which(Y>0)])
22   data.neg <- merge(X[which(X<0)],Y[which(Y<0)])
23   return(sum(data.pos[,1]<data.pos[,2])+sum(data.neg[,1]>data.neg[,2]))
24 }
25
26 Sukhatme.pvalue <- function(X,Y,nrep){
27   # Calculate the Sukhatme test statistics
28   n <- length(X)
29   stats <- Sukhatme.stats(X, Y)
30   # Simulate under the null
31   Sukhatme.sim <- c()
32   for (i in 1:nrep) {
33     Sukhatme.sim[i] <- Sukhatme.stats(rnorm(n,0,1), rnorm(n,0,1))
34   }
35   return(return(2*min(mean(Sukhatme.sim < stats), mean(Sukhatme.sim > stats))))
36 }
```

```

37
38 # Calculate the p-value of the statistics under the problem setting
39 set.seed(217)
40 X = rnorm(100); Y = rnorm(100,0,1.2); nrep = 500
41 moods.pvalue(X,Y,nrep)
42 Sukhatme.pvalue(X,Y,nrep)
43
44 ## Problem 1.2
45 # Case 1: False rejection when median is not 0
46 X = rnorm(100,1,1); Y = rnorm(100,1,1); nrep = 500
47 # Case 2: n is small
48 set.seed(217)
49 X = rnorm(10,0,1); Y = rnorm(10,0,5); nrep = 500
50 # Case 3: n is extremely large, Sukhatme brings larger computation cost
51 # and become less powerful than the Mood test
52
53 ## Problem 1.3
54 # Case 1: Skewed distribution
55 set.seed(217)
56 X = rnorm(50,1,0.5); Y = rgamma(50,0.5,0.5); nrep = 500

```

Both of the test failed to reject the null under the setting of Problem 1.1 with respectively p -value of 0.208 and 0.596. Scenario of Problem 1.2 and 1.3 are listed in codes. \square

Problem 2 Suppose X and Y follows the same continuous distribution and are independent. If the median of the distribution is 0, show that $\mathbb{P}(\{Y < X < 0\} \cup \{0 < X < Y\}) = \frac{1}{4}$.

Proof: Using the probability of conditional probability, it follows

$$\mathbb{P}(Y < X < 0) = \mathbb{P}(Y < X | X < 0, Y < 0) \mathbb{P}(Y < 0, X < 0) = \frac{1}{4},$$

where the last equality is based on the independency and property of median. Similarly, we have $\mathbb{P}(Y > X > 0) = \frac{1}{4}$ and the result follows.

□

Problem 3 Let $X_1, \dots, X_n \stackrel{iid}{\sim} N(0, 1) = F_X$ and $Y_1, \dots, Y_n \stackrel{iid}{\sim} N(\mu, 1) = F_Y$. Use simulation to answer the following: 1) Suppose $n = 100$ and $\mu = 0.1$, use the Wilcoxon Rank-Sum test and the Terry-Hoeffding test to test $H_0 : F_X = F_Y$ at level $\alpha = 0.05$. 2) Can you think of a scenario where the Wilcoxon Rank-Sum test is more powerful than the Terry-Hoeffding test? 3) Can you think of a scenario where the Terry-Hoeffding test is more powerful than the the Wilcoxon Rank-Sum test?

Solution: (Solution is not unique and the following code is based on the number of run)

```

1 TH.stats <- function(X,Y) {
2   X <- sort(X); Y <- sort(Y)
3   XY <- sort(c(X,Y))
4   rank.X <- match(X, XY)
5   weight <- normorder.expect(length(XY))
6   return(sum(weight[rank.X]))
7 }
8
9 TH.pvalue <- function(X,Y,nrep) {
10  # Calculate the Terry-Hoeffding Rank Sum test statistics
11  n <- length(X)
12  stats <- TH.stats(X, Y)
13  # Simulate under the null
14  TH.sim <- c()
15  for (i in 1:nrep) {
16    TH.sim[i] <- TH.stats(rnorm(n,0,1), rnorm(n,0,1))
17  }
18  return(2*min(mean(TH.sim < stats), mean(TH.sim > stats)))

```

```
19 }
20
21 # Calculate the p-value of the statistics under the problem setting
22 set.seed(217)
23 X = rnorm(100); Y = rnorm(100,0.1,1); nrep = 500
24 wilcoxsum.pvalue(X,Y,nrep)
25 TH.pvalue(X,Y,nrep)
26
27 # Problem 3.2
28 # Case 1: Samplings from some non-normal distributions
29 X = runif(15,0.5,1); Y = runif(15,-1,2); nrep = 500
30
31
32 # Problem 3.3
33 # Case 1: Terry-Hoeffding Rank Sum test is theoretically optimal
34 # under the normal assumption
35 X = rnorm(50,0,2); Y = rnorm(50,0.1,2); nrep = 500
36 }
```

Both of the test failed to reject the null under the setting of Problem 1.1 with respectively p -value of 0.808 and 0.772. Scenario of Problem 3.2 and 3.3 are listed in codes and you can read [Hodges Jr and Lehmann, 1960] for more technical details. \square

References

[Hodges Jr and Lehmann, 1960] Hodges Jr, J. and Lehmann, E. L. (1960). Comparison of the normal scores and wilcoxon tests. In *Proceedings 4th Berkeley Symposium*, volume 1, pages 307–317.