Solution for Problem 1

1.1

To get Kendall's
$$\tau$$
, we use the following test statistic:
$$T = \frac{\sum_{i=1}^{n} \sum_{j=1}^{n} U_{ij} V_{ij}}{\left[\left(\sum_{i=1}^{n} \sum_{j=1}^{n} U_{ij}^{2}\right)\left(\sum_{i=1}^{n} \sum_{j=1}^{n} V_{ij}^{2}\right)\right]^{1/2}} = \frac{\left[\left(\sum_{i=1}^{n} \sum_{j=1}^{n} U_{ij}^{2}\right)\left(\sum_{i=1}^{n} \sum_{j=1}^{n} V_{ij}^{2}\right)\right]^{1/2}}{\left[\left(\sum_{i=1}^{n} \sum_{j=1}^{n} U_{ij}^{2}\right)\left(\sum_{i=1}^{n} \sum_{j=1}^{n} V_{ij}^{2}\right)\right]^{1/2}}$$

$$\frac{\sum_{i=1}^{n}\sum_{j=1}^{n}U_{ij}V_{ij}}{n(n-1)}, \text{ where } U_{ij} = sgn(X_{j} - X_{i}), \ V_{ij} = sgn(Y_{j} - Y_{i}). \ Z = \frac{3\sqrt{n(n-1)}}{\sqrt{2(2n+5)}}T \sim N(0,1)$$

We can calculate Z statistic and compare it with the critical value $z_{\alpha/2} = \Phi^{-1} \left(1 - \frac{\alpha}{2} \right)$. If $|Z| \le z_{\alpha/2}$, we cannot reject the null hypothesis that X and Y are independent.

From 1,000 simulations, 93.8% of the simulations reach the conclusion that the null hypothesis cannot be rejected. Therefore, we cannot reject the null hypothesis that X and Y are independent using Kendall's τ .

1.2

To get Spearman's correlation, we use the following test statistic:
$$R=\frac{\sum_{i=1}^n(R_i-\bar{R})(S_i-\bar{S})}{\left[\left(\sum_{i=1}^n(R_i-R)^2\right)\left(\sum_{i=1}^n(S_i-\bar{S})^2\right)\right]^{1/2}}=1-\frac{6\sum_{i=1}^nD_i^2}{n(n^2-1)},$$
 where $R_i=rank(X_i),$ $S_i=rank(Y_i),$ $D_i=R_i-S_i.$
$$t=\frac{R\sqrt{n-2}}{1-R^2}\sim t_{n-2}$$

We can calculate t statistic and compare it with the critical value $t_{\alpha/2} = t_{n-2}^{-1} \left(1 - \frac{\alpha}{2}\right)$. If $|t| \le t_{\alpha/2}$, we cannot reject the null hypothesis that X and Y are independent.

From 1,000 simulations, 93.6% of the simulations reach the conclusion that the null hypothesis cannot be rejected. Therefore, we cannot reject the null hypothesis that X and Y are independent using Spearman's correlation.

Solution for Problem 2

We can use the same method from Problem 2 to test the null hypothesis using Kendall's τ and Spearman's correlation.

From 1,000 simulations, 94.2% of the simulations reach the conclusion that the null hypothesis cannot be rejected using Kendall's τ ; 93.9% of the simulations reach the conclusion that the null hypothesis cannot be rejected using Spearman's correlation. Therefore, we cannot reject the null hypothesis that X and Y are independent using Kendall's τ nor Spearman's correlation.

Kendall's τ is more powerful in this setting because X and Y are actually independent and Kendall's τ get more accurate results from simulations then Spearman's correlation.

Appendix

The following provides R code I used to complete this homework. You can also find the R code and output at https://thisiskunmeng.github.io/nonparametric/hw5.html

R code for Solution for Problem 1

```
#' rank function
rank = function(x){
    n = length(x)
    rank = c()
    for (i in 1:n){
        rank = c(rank, sum(x <= x[i]))
    return(rank)
#' Kendall's test
#' @return TRUE if the null hypothesis is not rejected
kendall <- function(x, y, a) {</pre>
    n <- length(x)</pre>
    u_ij <- c()
    v_ij <- c()</pre>
    for (i in 1:n) {
        for (j in 1:n) {
             u_ij <- c(u_ij, sign(x[j] - x[i]))
             v_ij <- c(v_ij, sign(y[j] - y[i]))</pre>
        }
    }
    t <- sum(u_ij * v_ij) / (n * (n - 1))
    z \leftarrow t * 3 * sqrt(n * (n - 1) / (2 * (2 * n + 5)))
    z_a \leftarrow qnorm(1 - a / 2)
    return(abs(z) <= z_a)</pre>
}
#' Spearman's test
#' @return TRUE if the null hypothesis is not rejected
spearman <- function(x, y, a) {</pre>
    n <- length(x)</pre>
    x_rank <- rank(x)</pre>
    y_rank <- rank(y)</pre>
    d <- x_rank - y_rank
    r \leftarrow 1 - 6 * sum(d^2) / (n * (n^2 - 1))
    t \leftarrow r * sqrt((n - 2) / (1 - r^2))
    t_a \leftarrow qt(1 - a / 2, n - 2)
    return(abs(t) <= t_a)</pre>
simulate_once <- function(seed) {</pre>
    set.seed(seed)
    n_pair <- 50
    x \leftarrow runif(n_pair, min = -1, max = 1)
    y \leftarrow rnorm(n_pair, mean = sqrt(1 - x^2), sd = 1)
    a <- 0.05
    return(c(kendall(x, y, a), spearman(x, y, a)))
}
n simulate <- 1000
kendall_result <- c()</pre>
spearman_result <- c()</pre>
for (i in 1:n_simulate) {
    result <- simulate_once(seed = i)</pre>
```

```
kendall_result <- c(kendall_result, result[1])</pre>
     spearman_result <- c(spearman_result, result[2])</pre>
print(paste("Kendall's test: ", sum(kendall_result) / n_simulate))
print(paste("Spearman's test: ", sum(spearman_result) / n_simulate))
# "Kendall's test: 0.938", "Spearman's test: 0.936"
R code for Solution for Problem 2
simulate_once <- function(seed) {</pre>
    set.seed(seed)
    n_pair <- 10
    x \leftarrow runif(n_pair, min = 1, max = 2)
    y \leftarrow rnorm(n_pair, mean = x / 2, sd = 1)
    a <- 0.05
    return(c(kendall(x, y, a), spearman(x, y, a)))
}
n_simulate <- 1000
kendall_result <- c()</pre>
spearman_result <- c()</pre>
for (i in 1:n_simulate) {
    result <- simulate_once(seed = i)</pre>
    kendall_result <- c(kendall_result, result[1])</pre>
    spearman_result <- c(spearman_result, result[2])</pre>
}
print(paste("Kendall's test: ", sum(kendall_result) / n_simulate))
print(paste("Spearman's test: ", sum(spearman_result) / n_simulate))
# "Kendall's test: 0.942", "Spearman's test: 0.939"
```