STATS 205: Homework Assignment 5

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Solution to Problem 1

We say that two observations X_1 and X_2 are independent of one another with respect to a collection of events A if

$$Pr\{X_1 \in A \text{ and } X_2 \in B\} = Pr\{X_1 \in A\} Pr\{X_2 \in B\}$$

where A and B are any two not necessarily distinct sets of outcomes belonging to A^3 .

- 2.2.1 Independent Observations; Permutation, Parametric, and Bootstrap Tests of Hypotheses; Good, Phillip I

In deciding whether your own observations are exchangeable and a permutation test applicable, the key question is the one we posed in the very first chapter: Under the null hypothesis of no differences among the various experimental or survey groups, can we exchange the labels on the observations without significantly affecting the results?

– 2.2.2 Exchangeable Observations; Permutation, Parametric, and Bootstrap Tests of Hypotheses; Good, Phillip I

Solution to Problem 2

```
cysticerci <- c(28.9, 32.8, 12.0, 9.9, 15.0, 38.0, 12.5, 36.5, 8.6, 26.8);cysticerci

## [1] 28.9 32.8 12.0 9.9 15.0 38.0 12.5 36.5 8.6 26.8

worms_reco <- c(1.0, 7.7, 7.3, 7.9, 1.1, 3.5, 18.9, 33.9, 28.6, 25.0); worms_reco
```

```
## [1] 1.0 7.7 7.3 7.9 1.1 3.5 18.9 33.9 28.6 25.0
```

The null hypothesis is that the mean weight of introduced cysticerci has no correlation with the mean weight of worms recovered. That is,

$$H_0: \tau = 0$$

The alternative hypothesis is that the mean weight of introduced cysticerci is *positively correlated with* the mean weight of worms recovered. That is,

$$H_A: \tau > 0$$

To test the null hypothesis against the alternative hypothesis, we will use the Kendall test, a distribution-free test for independence based on signs.

```
cor.test(x = cysticerci, y = worms_reco, method = "kendall", alt = "greater")
```

Kendall's rank correlation tau

##

```
## data: cysticerci and worms_reco
## T = 19, p-value = 0.7578
## alternative hypothesis: true tau is greater than 0
## sample estimates:
## tau
## -0.1555556
```

The p-value is 0.7578, which is not significant at the $\alpha = 0.05$ level. There is not enough evidence that the mean weight of introduced cysticerci is positively correlated with the mean weight of worms recovered.

Solution to Problem 3

```
cysticerci <- c(28.9, 32.8, 12.0, 9.9, 15.0, 38.0, 12.5, 36.5, 8.6, 26.8) worms_reco <- c(1.0, 7.7, 7.3, 7.9, 1.1, 3.5, 18.9, 33.9, 28.6, 25.0) cor.test(x = cysticerci, y = worms_reco, method = "kendall", alt = "greater") ## ## Kendall's rank correlation tau ## ## data: cysticerci and worms_reco ## T = 19, p-value = 0.7578 ## alternative hypothesis: true tau is greater than 0 ## sample estimates: ## tau ## -0.1555556 The estimate for \tau = -0.1555556.
```

Solution to Problem 4

Solution to Problem 5

```
cysticerci <- c(28.9, 32.8, 12.0, 9.9, 15.0, 38.0, 12.5, 36.5, 8.6, 26.8)
worms_reco <- c(1.0, 7.7, 7.3, 7.9, 1.1, 3.5, 18.9, 33.9, 28.6, 25.0)
```

The null hypothesis is that the mean weight of introduced cysticerci has no correlation with the mean weight of worms recovered. That is,

$$H_0: r_s < r_{s,\alpha}$$

The alternative hypothesis is that the mean weight of introduced cysticerci is *positively correlated with* the mean weight of worms recovered. That is,

$$H_A: r_s \geq r_{s,\alpha}$$

Otherwise, do not reject.

To test the null hypothesis against the alternative hypothesis, we will use the Spearman test, a distribution-free test for independence based on ranks.

```
# this method of performing the test was given in the textbook
library(SuppDists)
qSpearman(p = 0.05, r = 10)
```

```
## [1] -0.5393939
```

Since $r_{s,\alpha} = -0.5393939$, we will reject the null hypothesis only if $r_s \ge -0.5393939$.

Calculating r_s ,

```
cor(x = cysticerci, y = worms_reco, method = "spearman")
```

```
## [1] -0.2
```

Since $r_s = -0.2$ and $r_{s,\alpha} = -0.5393939$, the statement $r_s \ge r_{s,\alpha}$ is true. Thus, we reject the null hypothesis. There is sufficient evidence that the mean weight of introduced cysticerei is positively correlated with the mean weight of worms recovered.

NOTE: At this point, I tried to use cor.test() with method = "spearman" but I got a different result than I expected, and I'm not sure why. Maybe I'm interpreting the output incorrectly?

```
cor.test(x = cysticerci, y = worms_reco, method = "spearman", alternative = "greater")
##
```

```
##
## data: cysticerci and worms_reco
## S = 198, p-value = 0.72
## alternative hypothesis: true rho is greater than 0
## sample estimates:
```

Spearman's rank correlation rho

rho ## -0.2

##

The p-value is 0.72, which is not significant at the $\alpha = 0.05$ level. There is not enough evidence that the mean weight of introduced cysticerci is positively correlated with the mean weight of worms recovered.

Solution to Problem 6

```
x = c(0, 5000, 10000, 15000, 20000, 25000, 30000, 100000)

y = c(0.924, 0.988, 0.992, 1.118, 1.133, 1.145, 1.157, 1.357)
```

The null hypothesis is that the mean weight of introduced cysticerci has no correlation with the mean weight of worms recovered. That is,

$$H_0: \beta = \beta_0$$

$$H_0: \beta = 0$$

The alternative hypothesis is that the mean weight of introduced cysticerci is *positively correlated with* the mean weight of worms recovered. That is,

$$H_A: \beta > \beta_0$$

$$H_A: \beta > 0$$

To test the null hypothesis against the alternative hypothesis, we will use the Theil test, a distribution-free test for the slope of the regression line.

library(NSM3) ## Loading required package: combinat ## ## Attaching package: 'combinat' ## The following object is masked from 'package:utils': ## ## combn ## Loading required package: MASS ## Loading required package: partitions ## Loading required package: survival ## fANCOVA 0.5-1 loaded ## Registered S3 methods overwritten by 'ggplot2': ## method from ## [.quosures rlang ## c.quosures rlang print.quosures rlang # theil(x, y, alpha=0.05, beta.0=0, type = "u")theil.fit = theil (x, у, beta.0 = 0, slopes=TRUE, type = "1", doplot = FALSE) theil.fit ## Alternative: beta less than 0 ## C = 28, C.bar = 1, P = 1## beta.hat = 0## alpha.hat = 0.975 ## All slopes: ## i j S.ij ## 1 2 1.280000e-05 ## 1 3 6.800000e-06 ## 1 4 1.293333e-05 ## 1 5 1.045000e-05 ## 1 6 8.840000e-06 ## 1 7 7.766667e-06 ## 1 8 4.330000e-06 ## 2 3 8.000000e-07 ## 2 4 1.300000e-05

2 5 9.666667e-06 ## 2 6 7.850000e-06 ## 2 7 6.760000e-06 ## 2 8 3.884211e-06 ## 3 4 2.520000e-05 ## 3 5 1.410000e-05 ## 3 6 1.020000e-05 ## 3 7 8.250000e-06

```
## 3 8 4.055556e-06
## 4 5 3.00000e-06
## 4 6 2.700000e-06
## 4 7 2.600000e-06
## 4 8 2.811765e-06
## 5 6 2.400000e-06
## 5 7 2.400000e-06
## 5 8 2.800000e-06
## 6 7 2.400000e-06
## 6 8 2.826667e-06
## 7 8 2.857143e-06
##
##
## 1 - alpha = 0.95 lower bound for beta:
## 0, Inf
theil.output = theil(x,
 у,
 beta.0 = 0,
 slopes=TRUE,
 type = "t", doplot = FALSE, alpha = .05)
c(theil.output$L, theil.output$U)
## [1] 0 0
```