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 $\mathcal{A}$ 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 $\mathcal{A}^3$.

– 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

```
brain_weight = c(515, 286, 469, 410, 461, 436, 479, 198, 389, 262, 536); length(brain_weight)
```

```
## [1] 11
```

```
fiber_count  = c(32500, 26800, 11410, 14850, 23640, 23820, 29840, 21830, 24650, 22500, 26000); length(fi
```

```
## [1] 11
```

```
library(bootstrap)
theta.hat = cor(brain_weight, fiber_count); theta.hat
```

```
## [1] 0.1604644
```

```
library(partitions)
n = 1000
# allCompositions = compositions(n, n);allCompositions[,1:5]
# allCompositions.sub = allCompositions[, sample(1:dim(allCompositions)[2], size=1000, replace=FALSE)]

# draw.bootstrap.samples = function(df){
#   n = dim(df)[1]
#   ind = sample(n, replace = TRUE)
#   cor.bootstrap.replicate = cor(df[ind, "LSAT"], df[ind, "GPA"])
#   return(cor.bootstrap.replicate)
# }
# R = 1000
```

```
# theta.hat.star = replicate(R, draw.bootstrap.samples(law))
# # make a ggplot
# library(ggplot2)
# theta.hat.star.df = data.frame(theta.hat.star = theta.hat.star)
# ggplot(theta.hat.star.df) +
#   geom_density(aes(x = theta.hat.star, y = ..scaled..),
#     fill = "lightblue") +
#   geom_hline(yintercept=0, colour="white", size=1) +
#   theme_bw() +
#   ylab("density") +
#   xlab(bquote(hat(theta))) +
#   geom_vline(xintercept = theta.hat, col = "red")+
#   scale_y_continuous(expand = c(0,0))
```

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 $r_s \geq -0.5393939$.~~

~~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 $r_s \geq r_{s,\alpha}$ is *true*. Thus, we *reject* the null hypothesis. There is *sufficient* evidence that the mean weight of introduced cysticerci 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?~~

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

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

```r
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.

```r
library(NSM3)
```

```
## Loading required package: combinat
```

```
##
## Attaching package: 'combinat'
```

```
## The following object is masked from 'package:utils':
##
##     combn
```

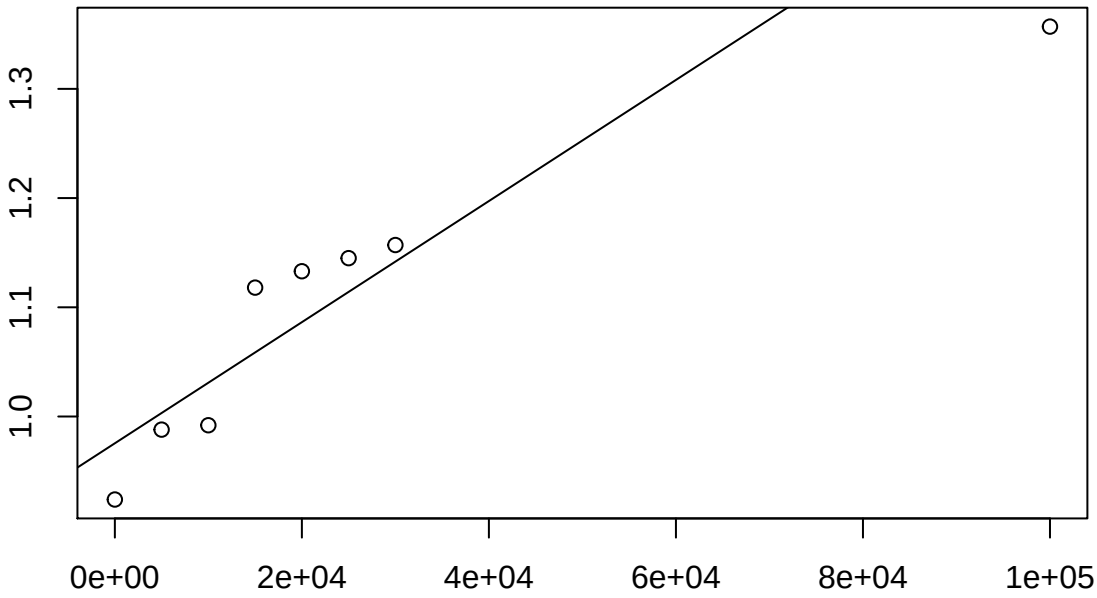
```
## Loading required package: MASS
```

```
## Loading required package: 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")
```




```
## Alternative: beta greater than 0
## C = 28, C.bar = 1, P = 0
## beta.hat = 0
## alpha.hat = 0.975
##
## 1 - alpha = 0.95 upper bound for beta:
## -Inf, 0
```

```
theil.fit = theil (x,
  y,
  beta.0 = 0 ,
  slopes=TRUE,
  type = "u",
  doplot = FALSE)
theil.fit
```

```
## Alternative: beta greater than 0
## C = 28, C.bar = 1, P = 0
## 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 3.000000e-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 upper bound for beta:
## -Inf, 0
```

```
theil.output = theil(x,
  y,
  beta.0 = 0,
  slopes=TRUE,
  type = "u", doplot = FALSE, alpha = .05)
c(theil.output$L, theil.output$U)
```

```
## [1] -Inf    0
```

TODO: Interpret these results correctly.

Solution to Problem 7

```
height = c(42.8, 63.5, 37.5, 39.5, 45.5, 38.5, 43.0, 22.5, 37.0, 23.5, 33.0, 58.0)
weight = c(40.0, 93.5, 35.5, 30.0, 52.0, 17.0, 38.5, 8.5, 33.0, 9.5, 21.0, 79.0)
heart_catheter_length = c(37.0, 49.5, 34.5, 36.0, 43.0, 28.0, 37.0, 20.0, 33.5, 30.5, 38.5, 47.0)

cor.test(x = height, y = heart_catheter_length, method = "pearson")
```

```
##
##  Pearson's product-moment correlation
##
## data:  height and heart_catheter_length
## t = 5.8936, df = 10, p-value = 0.0001524
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
##  0.6216270 0.9663721
```

```
## sample estimates:
##       cor
## 0.8811691
```

```
cor.test(x = weight, y = heart_catheter_length, method = "pearson")
```

```
##
##  Pearson's product-moment correlation
##
## data:  weight and heart_catheter_length
## t = 6.3033, df = 10, p-value = 8.871e-05
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
##  0.6568763 0.9700971
## sample estimates:
##       cor
## 0.8938226
```

From the Pearson correlation tests, there is *strong evidence* that, individually, height and weight contribute to the determination of heart catheter length.

```
library(Rfit)

r.01 <- rfit(heart_catheter_length ~ height)
f.01 <- rfit(heart_catheter_length ~ height + weight)
first_drop_test <- drop.test(f.01, r.01)
first_drop_test
```

```
##
## Drop in Dispersion Test
## F-Statistic     p-value
##     1.55202     0.24429
```

```
r.02 <- rfit(heart_catheter_length ~ weight)
second_drop_test <- drop.test(f.01, r.02)
second_drop_test
```

```
##
## Drop in Dispersion Test
## F-Statistic     p-value
##    0.014435    0.907007
```

However, based on the large p-values from the Drop in Dispersion tests, there is *not enough evidence* to suggest that height or weight contribute significantly over each other to the determination of heart catheter length.

Note

> Treating length of heart catheter as the *independent* variable, test for the importance of height and weight in *determining* the required catheter length.

If height and weight are the *determiners* of length of heart catheter, length of heart catheter must be the *dependent* variable.