Homework 7

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The homework for week seven is exercises 3, 4, 8, 9, and 10 on pages 155 and 156.

Question 3

Run cor.test() on the correlation between area and perm in the rock data set and interpret the results. Note that you will have to use the \$ accessor to get at each of the two variables (like this: rock\$area).

```
set.seed(3)
MyRock <- rock

RockCor <- cor.test(MyRock$area,MyRock$perm)
RockCor

##
## Pearson's product-moment correlation
##
## data: MyRock$area and MyRock$perm
## t = -2.9305, df = 46, p-value = 0.005254
## alternative hypothesis: true correlation is not equal to 0
## 95 percent confidence interval:
## -0.6118206 -0.1267915
## sample estimates:
## cor
## -0.396637</pre>
```

Interpret the Results

Make sure that you interpret both the confidence interval and the p-value that is generated by cor.test().

- The confidence interval does not pass 0
- The confidence interval shows that there is a 95% confidence that the correlation are somewhere between -0.6118 and -0.1267.
- Due to the P-Value being at 0.052, provides further confidence that the correlation could be near -0.396637.

Create a copy of the bfCorTest() custom function presented in this chapter. Don't forget to source it (meaning that you have to run the code that defines the function one time to make R aware of it). Conduct a Bayesian analysis of the correlation between area and perm in the rock data set.

```
bfCorTest <- function (x,y) # Get r from BayesFactor
{
    zx <- scale(x) # Standardize X
    zy <- scale(y) # Standardize Y
    zData <- data.frame(x=zx,rhoNot0=zy) # Put in a data frame
    bfOut <- generalTestBF(x ~ rhoNot0, data=zData) # linear coefficient
    mcmcOut <- posterior(bfOut,iterations=10000) # posterior samples
    print(summary(mcmcOut[,"rhoNot0"])) # Show the HDI for r
    return(bfOut) # Return Bayes factor object
}
bfCorTest(MyRock$area,MyRock$perm)</pre>
```

```
##
## Iterations = 1:10000
## Thinning interval = 1
## Number of chains = 1
## Sample size per chain = 10000
##
## 1. Empirical mean and standard deviation for each variable,
##
      plus standard error of the mean:
##
                                        Naive SE Time-series SE
##
             Mean
                               SD
##
        -0.347974
                        0.137807
                                        0.001378
                                                        0.001552
##
## 2. Quantiles for each variable:
##
##
       2.5%
                 25%
                           50%
                                    75%
                                           97.5%
## -0.62128 -0.43917 -0.34554 -0.25479 -0.08356
## Bayes factor analysis
## [1] rhoNot0 : 8.072781 ±0%
## Against denominator:
     Intercept only
##
## ---
## Bayes factor type: BFlinearModel, JZS
```

Interpret the Results

- The point estimate is close to the cor.test at -0.347974.
- The 95% HDI is somewhere between -0.6118 and -0.1267 and does not cross 0.
- Due to the rhoNot0 at 8.07 provides further evidence that the odds are in favor of the alternative hypothesis.

- In order to make 8.07 more intuitive, inverting the value like the following 1/8.0727 comes out to be 0.12:1 which is not great odds in favor of the null hypothesis.
- Any odds ratio weaker than 3:1 is not worth mentioning.

Not unexpectedly, there is a data set in R that contains these data. The data set is called UCBAdmissions and you can access the department mentioned above like this: UCBAdmissions[, ,1]. Make sure you put two commas before the 1: this is a three dimensional contingency table that we are subsetting down to two dimensions. Run chisq.test() on this subset of the data set and make sense of the results.

```
MyUCB <- UCBAdmissions[ , ,1]</pre>
MyUCB
             Gender
##
## Admit
              Male Female
     Admitted 512
                        89
##
##
     Rejected 313
                        19
chisq.test(MyUCB)
##
   Pearson's Chi-squared test with Yates' continuity correction
##
##
## data: MyUCB
## X-squared = 16.372, df = 1, p-value = 5.205e-05
```

Interpret the Results

- In this case the chi-squared test is looking at a 2x2 table in relation to admittance of students at UC Berkeley by gender.
- The the Chi-Squared is 16.372 with 1 degree of freedom.
- Degrees of freedom are the values that have the freedom to vary in the admittance data.
- Due to having a 2 X 2 table on only 1 variable can vary or could be manipulated.
- The p-value of 5.205e-05 is extremely low therefore Male and Females are not independent and specifically UC Berkley would reject a male applicants over woman.

Use contingencyTableBF() to conduct a Bayes factor analysis on the UCB admissions data.

```
BayesUCB <- contingencyTableBF(MyUCB, sampleType = 'poisson')
BayesUCB</pre>
```

```
## Bayes factor analysis
## -----
## [1] Non-indep. (a=1) : 1111.64 ±0%
##
## Against denominator:
## Null, independence, a = 1
## ---
## Bayes factor type: BFcontingencyTable, poisson
```

Interpret the Results

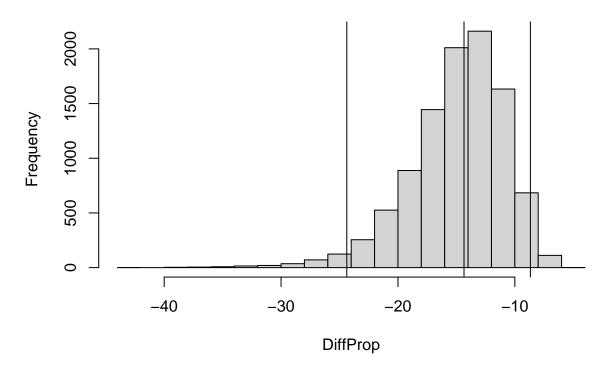
Report and interpret the Bayes factor.

- The Non-independent number of 1111.64 + or 0% is the Bayes factor and is help to measure if the odds are favor of the alternative Hypothesis.
- Due to having 1111.64 the odds are in the favor and further help prove that the Chi-Square is true.
- Typically anything over 150:1 is illustrates very strong evidence in favor of the hypothesis.

Using the UCBA data, run contingencyTableBF() with posterior sampling. Use the results to calculate a 95% HDI of the difference in proportions between the columns.

```
BayesUCB_Post <- contingencyTableBF(MyUCB, sampleType = 'poisson', posterior = TRUE, iterations = 100</pre>
summary(BayesUCB_Post)
##
## Iterations = 1:10000
## Thinning interval = 1
## Number of chains = 1
## Sample size per chain = 10000
## 1. Empirical mean and standard deviation for each variable,
##
      plus standard error of the mean:
##
##
                 Mean
                          SD Naive SE Time-series SE
## lambda[1,1] 511.11 22.553 0.22553
                                             0.22961
## lambda[2,1] 312.55 17.814 0.17814
                                             0.17814
## lambda[1,2] 89.67 9.384 0.09384
                                             0.09384
## lambda[2,2] 19.86 4.475 0.04475
                                             0.04475
##
## 2. Quantiles for each variable:
##
                 2.5%
                         25%
                                50%
                                       75% 97.5%
## lambda[1,1] 467.72 495.75 510.72 525.81 556.74
## lambda[2,1] 278.87 300.51 312.10 324.52 348.25
## lambda[1,2] 72.14 83.13 89.36 95.93 109.07
## lambda[2,2] 12.07 16.76 19.51 22.61 29.61
MaleProp <- BayesUCB_Post[,'lambda[1,1]']/BayesUCB_Post[,'lambda[2,1]']</pre>
FemaleProp <- BayesUCB_Post[,'lambda[2,1]']/BayesUCB_Post[,'lambda[2,2]']
DiffProp <- MaleProp - FemaleProp</pre>
mean(DiffProp)
## [1] -14.93356
hist(DiffProp)
abline(v=quantile(DiffProp,c(0.025)), col='black')
abline(v=quantile(DiffProp,c(0.975)), col='black')
abline(v=quantile(DiffProp,c(0.500)), col='black')
```

Histogram of DiffProp



Interpret the Results

- When looking at the HDI, each represents males admitted and rejected; and females admitted and rejected.
- $\bullet\,$ At 95% confidence males accepted range is from 468 to 557.
- \bullet Alternatively at 95% confidence the Males rejected was 279 to 348.
- $\bullet\,$ The female accepted between a range of 72 and 109 at 95% confidence.
- At 95% confidence the female rejections numbers were between 12 to 30.