

Solution for the R part of Lab 3

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```
library(foreign)
library(car)
load("GSS.Rdata")
# summary(GSS) #commented out for the report
```

1. Chi-squared test

```
##      married      widowed      divorced      separated never married
##      795          165          213          40          286
##      NA
##      1
```

```
GSS$marital[GSS$marital=="NA"] = NA
GSS$marital[GSS$marital=="NA"]
```

```
## [1] <NA>
## Levels: married widowed divorced separated never married NA
```

```
summary(GSS$politics)
```

```
##      Liberal      Tend Lib      Moderate      Tend Cons      Conservative
##      193          193          527          248          282
##      NA's
##      57
```

```
table(GSS$marital, GSS$politics)
```

```
##
##      Liberal Tend Lib Moderate Tend Cons Conservative
## married      93      92      271      140      173
## widowed      15      16       57       24       37
## divorced     22      36       79       38       29
## separated      7       3       22        6        1
## never married  55      46       98       40       42
## NA            0       0        0        0        0
```

```
cs = chisq.test(GSS$marital, GSS$politics)
```

```
cs$statistic
```

```
## X-squared
## 44.2255
```

```
# cs$stdres  
# cs$expected
```

```
library(lsr)
```

```
## Warning: package 'lsr' was built under R version 3.1.3
```

```
cramersV(cs$observed)
```

```
## [1] 0.08756363
```

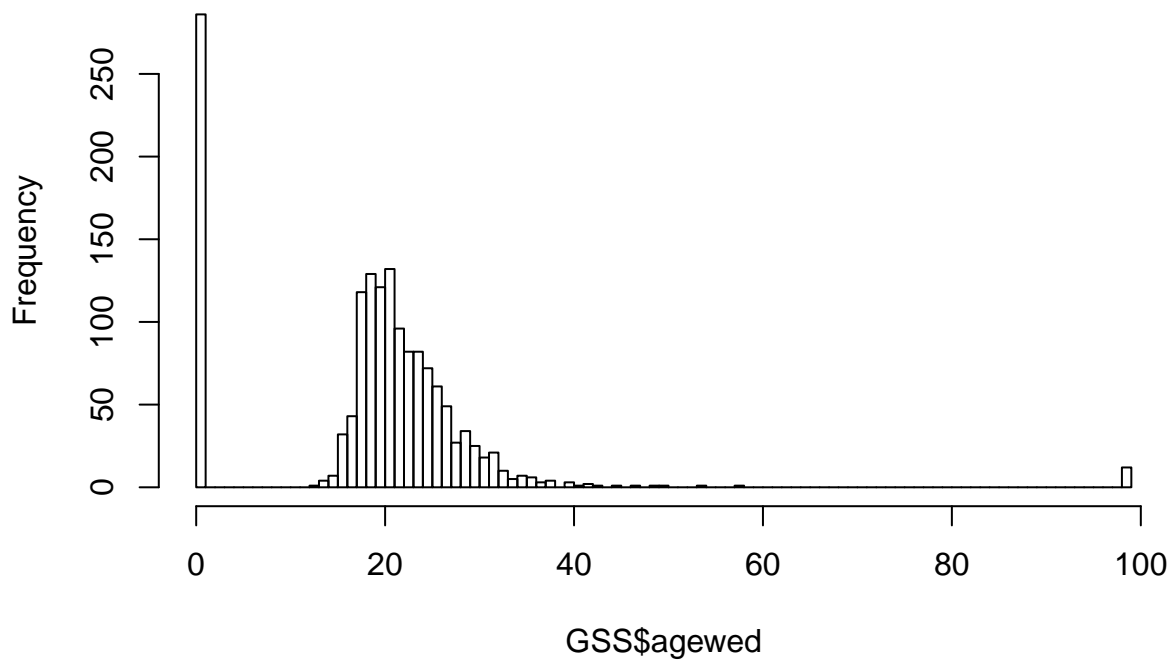
2. Correlation

```
# Conduct an analysis to examine the  
# relationship between agewed and tvhours.  
summary(GSS$agewed)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.  
##      0.00   18.00   21.00   19.06   24.00   99.00
```

```
hist(GSS$agewed, breaks= 100)
```

Histogram of GSS\$agewed



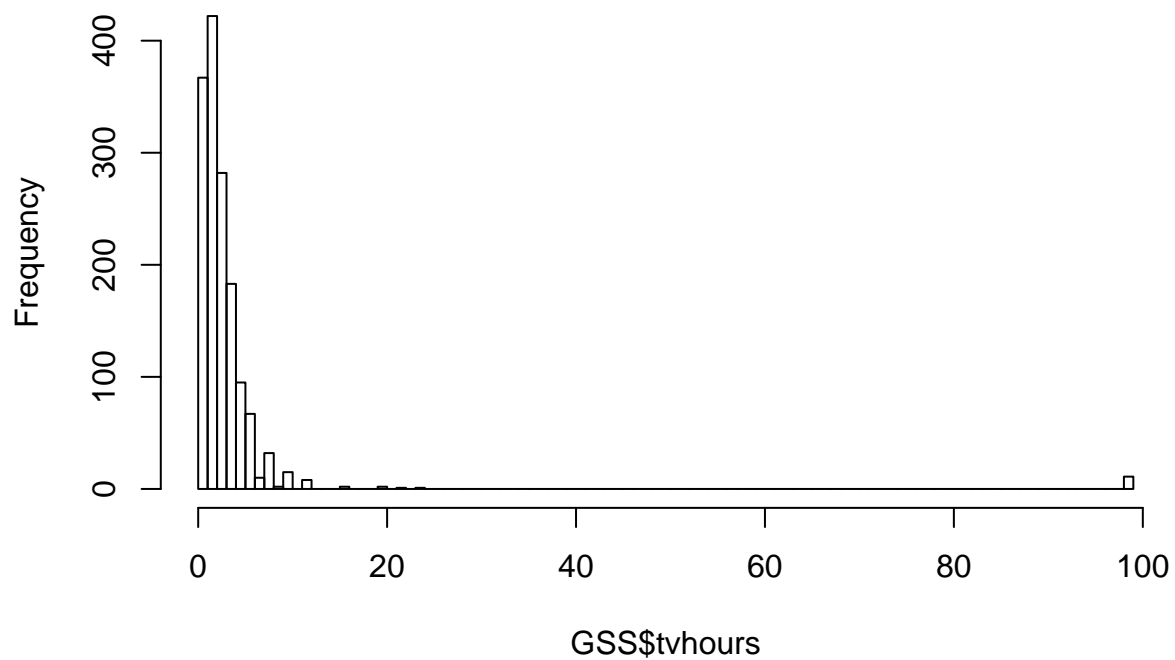
```
GSS$agewed[GSS$agewed==0] = NA
GSS$agewed[GSS$agewed==99] = NA
```

```
# tv hours
summary(GSS$tvhours)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##    0.000   2.000   2.000   3.605   4.000  99.000
```

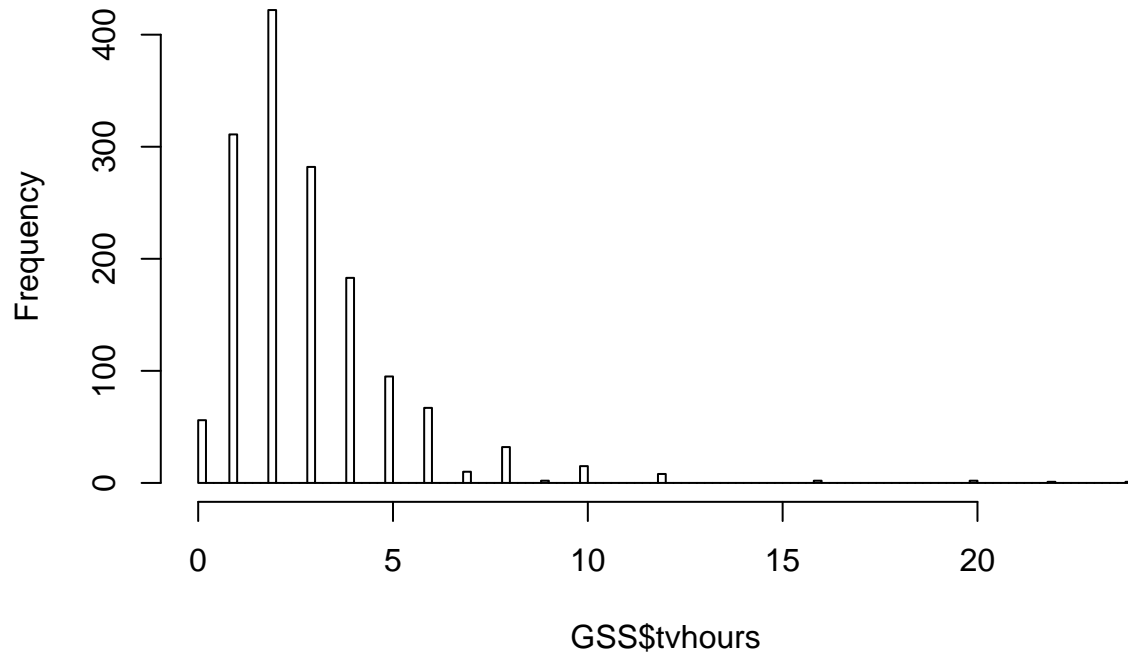
```
hist(GSS$tvhours, breaks= 100)
```

Histogram of GSS\$tvhours

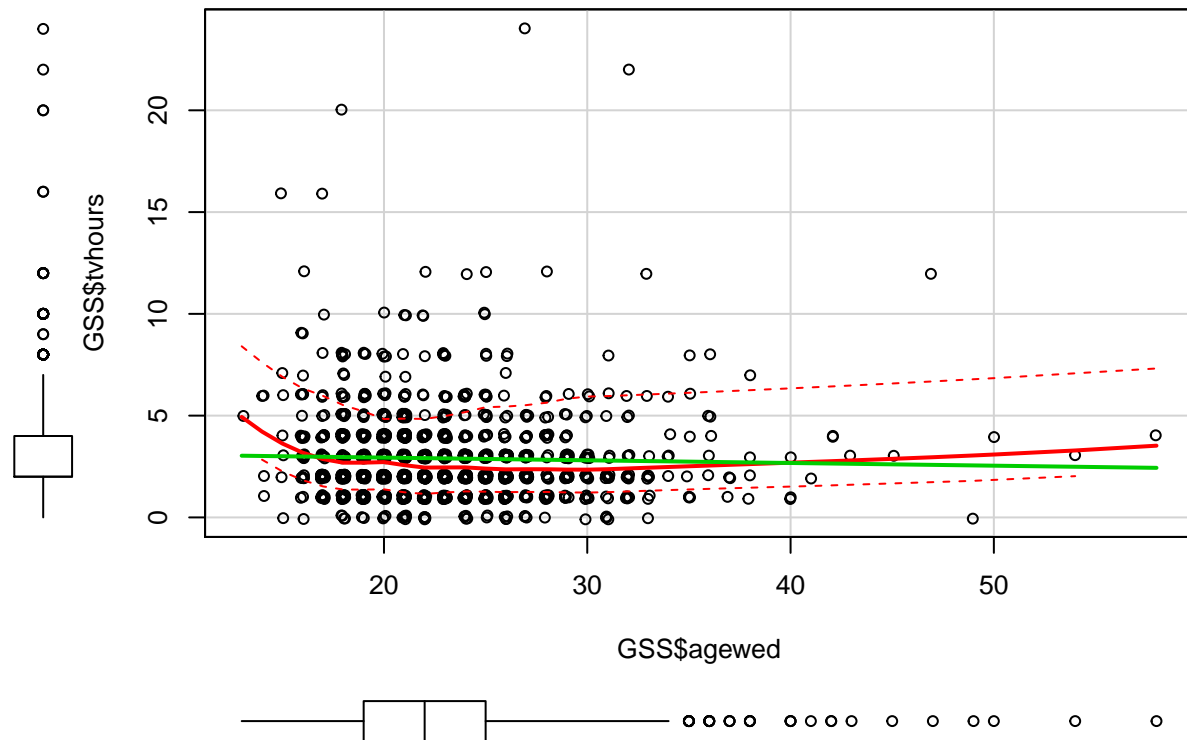


```
GSS$tvhours[GSS$tvhours==99] = NA
hist(GSS$tvhours, breaks= 100)
```

Histogram of GSS\$tvhours



```
scatterplot(GSS$agewed, GSS$tvhours, jitter = list(x=.5,y=.5), lwd=2)
```



```
ct = cor.test(GSS$agewed, GSS$tvhours)
cat ( 'r = ', ct$estimate, '....' , 'p-value=' , ct$p.value )
```

```
## r = -0.02996096 .... p-value= 0.3009361
```

3. T-test

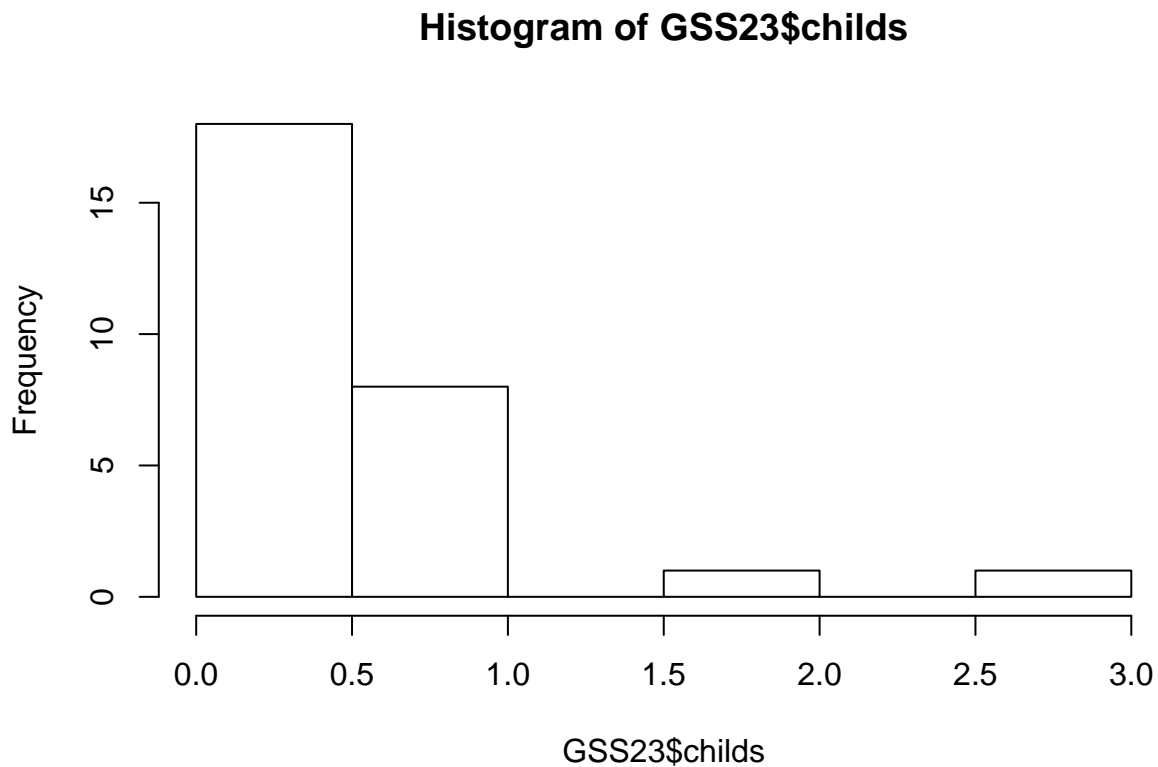
```
GSS23=GSS[GSS$age==23, ]  
# summary(GSS23)  
  
summary(GSS23$marital)
```

```
##      married      widowed      divorced      separated never married  
##          8           0           0           0             20  
##         NA  
##          0
```

```
GSS23$married = ifelse(GSS23$marital == "married", 1, 0)  
  
# percent of married 23-year olds  
mean( as.numeric(GSS23$married, na.rm=T) )
```

```
## [1] 0.2857143
```

```
hist(GSS23$chlds)
```



```
wt = wilcox.test(childs ~ married, data = GSS23)
```

```
## Warning in wilcox.test.default(x = c(0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, :  
## cannot compute exact p-value with ties
```

```
wt$statistic
```

```
## W  
## 19
```

```
wt$p.value
```

```
## [1] 0.0002656205
```

```
# Using the formula in the book (ch.15)  
# for effect size out of wilcox.test
```

```
rFromWilcox<-function(wilcoxModel, N){  
  z<- qnorm(wilcoxModel$p.value/2)  
  r<- z/ sqrt(N)  
  cat(wilcoxModel$data.name, "Effect Size, r = ", r,"\n")  
  r  
}
```

```
N = dim(GSS23)[1]  
rFW = rFromWilcox(wt,N)
```

```
## childs by married Effect Size, r = -0.6891632
```

Something to note above Field's function is that the way he calculates the z-score from the p-value, we always get a negative z-score, so the result is always going to be negative. Here, we know for a fact that the effect is positive; for example, see `t.test(childs ~ married, data = GSS23)` to see why.). But we really just care about the size of `rFromWilcox` not its sign. Or, just square it be happy.

```
cat('r = ', abs(rFW) , '..... r^2 = ',rFW^2)
```

```
## r = 0.6891632 ..... r^2 = 0.4749459
```

The above result is what we are looking for. But one could also do `cohens_d` or a regular effect size correlation while ignoring that we have been asked to do a wilcoxon rank-sum test.

```
cor.test(GSS23$childs, GSS23$married)
```

```
##  
## Pearson's product-moment correlation  
##  
## data: GSS23$childs and GSS23$married  
## t = 4.7257, df = 26, p-value = 6.934e-05  
## alternative hypothesis: true correlation is not equal to 0
```

```
## 95 percent confidence interval:
## 0.4108605 0.8398404
## sample estimates:
##      cor
## 0.6797444
```

```
cohens_d <- function(x,y)
{
  # this function takes two vectors as input and compares their means
  #lets compute the pooled standard error
  lx = length(subset(x,!is.na(x)))
  ly = length(subset(y,!is.na(y)))
  #numerator of the pooled variance
  num = (lx-1)*var(x,na.rm=T) + (ly-1)*var(y,na.rm=T)
  pooled_var = num / (lx + ly -2)
  pooled_sd = sqrt(pooled_var)
  # cohen's d
  cd = abs(mean(x,na.rm=T)-mean(y,na.rm=T))/pooled_sd
  return(cd)
}
```

```
with(GSS23, cohens_d(childs[married==0] , childs[married==1]) )
```

```
## [1] 1.976885
```

4. ANOVA

```
#examine relationship between relig and agewed
summary(GSS$relig)
```

```
## Protestant   Catholic    Jewish      None      Other      DK
##          953         333         31        140         35         1
##           NA
##           7
```

```
GSS$relig[GSS$relig == "NA"] = NA
GSS$relig[GSS$relig == "DK"] = NA
GSS$relig = factor(GSS$relig)
summary(GSS$relig)
```

```
## Protestant   Catholic    Jewish      None      Other      NA's
##          953         333         31        140         35         8
```

```
aov_model = aov(agewed ~ relig, data = GSS)
summary.lm(aov_model)
```

```
##
## Call:
## aov(formula = agewed ~ relig, data = GSS)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -9.634 -3.253 -1.253  2.366 35.747
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   22.2529    0.1772 125.553 < 2e-16 ***
## religCatholic  1.3811    0.3531   3.911 9.71e-05 ***
## religJewish   3.3993    1.0518   3.232 0.00126 **
## religNone     1.0629    0.5400   1.968 0.04927 *
## religOther    3.1702    0.9911   3.199 0.00142 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 4.972 on 1191 degrees of freedom
## (304 observations deleted due to missingness)
## Multiple R-squared:  0.02679,    Adjusted R-squared:  0.02352
## F-statistic: 8.197 on 4 and 1191 DF,  p-value: 1.606e-06
```

```
summary(aov_model)
```

```
##              Df Sum Sq Mean Sq F value    Pr(>F)
## relig           4      811   202.64    8.197 1.61e-06 ***
## Residuals     1191   29444    24.72
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## 304 observations deleted due to missingness
```

```
# remember that eta^2
# in the anova lingo is: R^2
eta.sqr = summary.lm(aov_model)$r.squared
eta = sqrt(eta.sqr)
cat( ' eta^2 = ',eta.sqr ,'\n eta = ' ,sqrt(eta))
```

```
## eta^2 = 0.02679156
## eta = 0.4045754
```

```
omega.sq <- function(aovm){
  sum_stats <- summary(aovm)[[1]]
  SSm <- sum_stats[["Sum Sq"]][1]
  SSr <- sum_stats[["Sum Sq"]][2]
  DFm <- sum_stats[["Df"]][1]
  MSr <- sum_stats[["Mean Sq"]][2]
  SSt <- SSm + SSr
  `w^2` <- (SSm-DFm*MSr)/(SSt+MSr)
  return(`w^2`)
}
```



```
omega.sqr = omega.sq(aov_model)
cat( ' omega^2 = ',omega.sqr ,'\n omega = ' ,sqrt(omega.sqr))
```

```
## omega^2 = 0.02350381
## omega = 0.1533095
```

```
pairwise.t.test(GSS$agewed, GSS$relig, p.adjust.method="bonferroni")
```

```
##
## Pairwise comparisons using t tests with pooled SD
##
## data: GSS$agewed and GSS$relig
##
##      Protestant Catholic Jewish  None
## Catholic 0.00097      -      -      -
## Jewish   0.01263    0.62106  -      -
## None     0.49274    1.00000 0.43398 -
## Other    0.01417    0.80224 1.00000 0.55751
##
## P value adjustment method: bonferroni
```

```
# significant comparisons: Catholic-Protestant (p=.00097), Jewish-Protestant (p=.01263),
# Other-Protestant (p=0.01417)
```