

Assignment 3

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October 11, 2016

Grade: 48/50

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part 1

```
set.seed(100)

loops <- 1e3
exp.pvals <- numeric(loops)
for(i in seq(loops)) {
  group <- rnorm(100, 60, 20)
  assign <- sample(c(0,1), 100, replace = TRUE)
  exp <- data.frame(group, assign)
  for(x in 1:100) {
    if(exp[x,'assign'] == 1) {
      exp[x,1] <- exp[x,1] +5
    }
    x <- x+1
  }
  mod.sim <- lm(group~assign, data = exp)
  summ <- summary(mod.sim)
  coefs <- coef(summ)
  exp.pvals[i] <- coefs[2,4]
  i <- i+1
}

sum(exp.pvals<=0.05)/loops
```

```
## [1] 0.245
```

The power for 100 samples is 0.231.

part 2

```
set.seed(1000)
loops_2 <- 1e3
exp.pvals_2 <- numeric(loops_2)
for(j in seq(loops_2)) {
  group_2 <- rnorm(1000, 60, 20)
  assign_2 <- sample(c(0,1), 1000, replace = TRUE)
  exp_2 <- data.frame(group_2, assign_2)
  for(y in 1:1000) {
```

```

    if(exp_2[y,'assign_2'] == 1) {
      exp_2[y,1] <- exp_2[y,1] +5
    }

  }
  mod.sim_2 <- lm(group_2~assign_2, data = exp_2)
  summ_2 <- summary(mod.sim_2)
  coefs_2 <- coef(summ_2)
  exp.pvals_2[j] <- coefs_2[2,4]
}

sum(exp.pvals_2<=0.05)/loops_2

```

```
## [1] 0.98
```

The power for 1000 samples is 0.98.

3

The football-values data set is stored in data frame `football`, which is then converted into `football` with the first two columns of the original data set removed.

```

football <- read.csv('proj_wr16.csv')
football <- football[,-1:-2]

```

part 1

The correlation matrix of `football` is shown using:

```

cor(football)

##          rush_att  rush_yds  rush_tds  rec_att  rec_yds  rec_tds
## rush_att 1.0000000 0.9906030 0.88608205 0.19706851 0.14473723 0.13548999
## rush_yds 0.9906030 1.0000000 0.91252627 0.18745520 0.13765791 0.12772327
## rush_tds 0.8860820 0.9125263 1.00000000 0.06914613 0.03114206 0.03163468
## rec_att  0.1970685 0.1874552 0.06914613 1.00000000 0.99002712 0.96757796
## rec_yds  0.1447372 0.1376579 0.03114206 0.99002712 1.00000000 0.98209522
## rec_tds  0.1354900 0.1277233 0.03163468 0.96757796 0.98209522 1.00000000
## fumbles  0.1844220 0.1881021 0.10845675 0.43577978 0.40349289 0.35852435
## fpts     0.1766540 0.1698501 0.06567865 0.98754942 0.99760259 0.99058639
##          fumbles      fpts
## rush_att 0.1844220 0.17665405
## rush_yds 0.1881021 0.16985010
## rush_tds 0.1084568 0.06567865
## rec_att  0.4357798 0.98754942
## rec_yds  0.4034929 0.99760259
## rec_tds  0.3585244 0.99058639
## fumbles  1.0000000 0.38269698
## fpts     0.3826970 1.00000000

```

part 2

To make a data set with a similar correlation structure, we use `mvrnorm` since the football data set consists of multiple variables and corresponding means and covariances. The resulting correlation matrix of the simulated data is contained in `keep.i`.

```
library(MASS)
rho <- cor(football)
vcov <- var(football)
means <- colMeans(football)

keep.i <- 0
loops <- 1e4
for(i in seq(loops)) {
  football.sim <- mvrnorm(30, mu=means, Sigma=vcov)
  keep.i <- keep.i + cor(football.sim)/loops
}

keep.i
```

```
##      rush_att rush_yds rush_tds rec_att rec_yds rec_tds
## rush_att 1.0000000 0.9902341 0.88241167 0.19362908 0.1418744 0.13289492
## rush_yds 0.9902341 1.0000000 0.90964176 0.18401614 0.1348112 0.12530075
## rush_tds 0.8824117 0.9096418 1.00000000 0.06921105 0.0315649 0.03213981
## rec_att  0.1936291 0.1840161 0.06921105 1.00000000 0.9896449 0.96653342
## rec_yds  0.1418744 0.1348112 0.03156490 0.98964486 1.0000000 0.98149861
## rec_tds  0.1328949 0.1253007 0.03213981 0.96653342 0.9814986 1.00000000
## fumbles  0.1827721 0.1861464 0.10810836 0.42843163 0.3965239 0.35290690
## fpts     0.1732943 0.1665474 0.06553721 0.98711766 0.9975202 0.99026166
##      fumbles      fpts
## rush_att 0.1827721 0.17329432
## rush_yds 0.1861464 0.16654738
## rush_tds 0.1081084 0.06553721
## rec_att  0.4284316 0.98711766
## rec_yds  0.3965239 0.99752024
## rec_tds  0.3529069 0.99026166
## fumbles  1.0000000 0.37620335
## fpts     0.3762033 1.00000000
```

part 3

The procedure for making a simulated data set with the SAME correlation structure as the football data set is the same, except we include the argument ‘`empirical`’ and set it equal to `TRUE` to show that the means and variances we use to create our simulated data are empirical means and variances rather than true means and variances.

```
keep.j <- 0
loops <- 1e4
for(j in seq(loops)) {
  football.sim <- mvrnorm(30, mu=means, Sigma=vcov, empirical = TRUE)
  keep.j <- keep.j + cor(football.sim)/loops
}

keep.j
```

```

##          rush_att  rush_yds  rush_tds  rec_att  rec_yds  rec_tds
## rush_att 1.0000000 0.9906030 0.88608205 0.19706851 0.14473723 0.13548999
## rush_yds 0.9906030 1.0000000 0.91252627 0.18745520 0.13765791 0.12772327
## rush_tds 0.8860820 0.9125263 1.00000000 0.06914613 0.03114206 0.03163468
## rec_att  0.1970685 0.1874552 0.06914613 1.00000000 0.99002712 0.96757796
## rec_yds  0.1447372 0.1376579 0.03114206 0.99002712 1.00000000 0.98209522
## rec_tds  0.1354900 0.1277233 0.03163468 0.96757796 0.98209522 1.00000000
## fumbles  0.1844220 0.1881021 0.10845675 0.43577978 0.40349289 0.35852435
## fpts     0.1766540 0.1698501 0.06567865 0.98754942 0.99760259 0.99058639
##          fumbles      fpts
## rush_att 0.1844220 0.17665405
## rush_yds 0.1881021 0.16985010
## rush_tds 0.1084568 0.06567865
## rec_att  0.4357798 0.98754942
## rec_yds  0.4034929 0.99760259
## rec_tds  0.3585244 0.99058639
## fumbles  1.0000000 0.38269698
## fpts     0.3826970 1.00000000

```

JC Grading -2 Looking for a single generated dataset.

Question 4

part 1

$$P(B) = \sum_j P(B|A_j)P(A_j), \Rightarrow P(A_i|B) = \frac{P(B|A_i)P(A_i)}{\sum_j P(B|A_j)P(A_j)} \quad (1)$$

part 2

$$\hat{f}(\zeta) = \int_{-\infty}^{\infty} f(x)e^{-2\pi i x \zeta} dx \quad (2)$$

part 3

$$\mathbf{J} = \frac{d\mathbf{f}}{d\mathbf{x}} = \left[\frac{\partial \mathbf{f}}{\partial x_1} \cdots \frac{\partial \mathbf{f}}{\partial x_n} \right] = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \cdots & \frac{\partial f_1}{\partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial f_m}{\partial x_1} & \cdots & \frac{\partial f_m}{\partial x_n} \end{bmatrix} \quad (3)$$