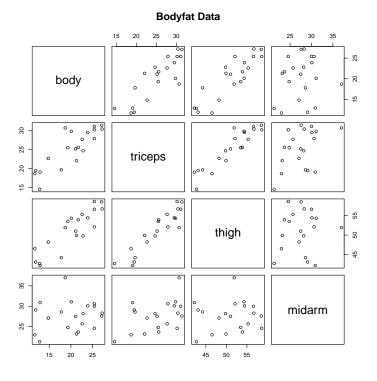
Stat 5100 Handout 2.6.1 - R: Inference with Multiple Predictors Stat 5100: Dr. Bean

Example: (Table 7.1) Study seeks to relate (in females) amount of body fat (Y) to triceps skinfold thickness (X_1) , thigh circumference (X_2) , and midarm circumference (X_3) . Amount of body fat is expensive to measure, requiring immersion of person in water. This expense motivates the desire for a predictive model based on these inexpensive predictors.

- Q1: Do thigh and midarm both have no effect on body fat when triceps is in the model?
- Q2: Do the relationships among the predictors cause any problems in the fitted model?

```
# Make output easier to read by disabling scientific notation
options(scipen = 999)
# Input data and take a look at the first few observations
library(stat5100)
data(bodyfat)
head(bodyfat)
    triceps thigh midarm body
## 1 19.5 43.1 29.1 11.9
       24.7 49.8 28.2 22.8
## 3
     30.7 51.9 37.0 18.7
       29.8 54.3 31.1 20.1
## 4
## 5
     19.1 42.2 30.9 12.9
       25.6 53.9 23.7 21.7
## 6
# Look at the correlation matrix
cor(bodyfat)
            triceps thigh
                                midarm
## triceps 1.0000000 0.9238425 0.4577772 0.8432654
## thigh 0.9238425 1.0000000 0.0846675 0.8780896
## midarm 0.4577772 0.0846675 1.0000000 0.1424440
## body 0.8432654 0.8780896 0.1424440 1.0000000
# Look at the scatterplot
pairs( ~ body + triceps + thigh + midarm, data = bodyfat,
      main = "Bodyfat Data")
```



Question 1: Test whether thigh and midarm BOTH have no effect on body when triceps is in the model

```
bodyfat_lm_full <- lm(body ~ triceps + thigh + midarm, data = bodyfat)
anova(bodyfat_lm_full)
## Analysis of Variance Table
##
## Response: body
            Df Sum Sq Mean Sq F value
##
                                          Pr(>F)
             1 352.27 352.27 57.2768 0.000001131 ***
## triceps
## thigh
             1 33.17
                      33.17 5.3931
                                         0.03373 *
## midarm
            1 11.55
                       11.55 1.8773
                                         0.18956
## Residuals 16 98.40
                        6.15
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
bodyfat_lm_reduced <- lm(body ~ triceps, data = bodyfat)</pre>
anova(bodyfat_lm_reduced)
## Analysis of Variance Table
##
## Response: body
            Df Sum Sq Mean Sq F value
                                          Pr(>F)
           1 352.27 352.27 44.305 0.000003024 ***
## triceps
## Residuals 18 143.12
                         7.95
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

Perform the subset F-test by hand

```
# (All these values are grabbed from the ANOVA tables above)
SSE_Reduced <- 143.12
                                    # Sum of squared error for reduced model
SSE_Full <- 98.40
                                    # Sum of squared error for full model
MSE_Full <- 6.15
                                    # Mean square error for full model
MSR <- (SSE_Reduced - SSE_Full) / 2 # Mean square reduction
F_statistic <- MSR / MSE_Full
F_statistic
## [1] 3.635772
# The F-statistic above follows a F(2,16) distribution (16 denominator degrees
# of freedom because MSE_Full is calculated by SSE_Full / 16)
pvalue <- pf(F_statistic, 2, 16, lower.tail = FALSE)</pre>
pvalue
## [1] 0.04992961
```

Perform the subset F-test automatically

```
# This is a test for the null hypothesis that thigh = midarm= 0
bodyfat_lm_full <- lm(body ~ triceps + thigh + midarm, data = bodyfat)
bodyfat_lm_reduced <- lm(body ~ triceps, data = bodyfat)</pre>
anova(bodyfat_lm_full, bodyfat_lm_reduced)
## Analysis of Variance Table
## Model 1: body ~ triceps + thigh + midarm
## Model 2: body ~ triceps
## Res.Df
              RSS Df Sum of Sq F Pr(>F)
## 1 16 98.405
## 2
       18 143.120 -2 -44.715 3.6352 0.04995 *
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
# Notice here that we get the same F-statistic and p-value as we did when
# we tested by hand.
```

Question 2: Investigate the effect of relationships among predictors

```
# Standardize the variables and create a standardized regression model
bodyfat_standardized <- data.frame(scale(bodyfat))
bodyfat_s_lm<- lm(body ~ triceps + thigh + midarm, data = bodyfat_standardized)
summary(bodyfat_s_lm)

##
## Call:
## lm(formula = body ~ triceps + thigh + midarm, data = bodyfat_standardized)
##
## Residuals:
## Min 1Q Median 3Q Max</pre>
```

```
## -0.72976 -0.31552 0.07683 0.28702 0.80837
## Coefficients:
                                        Std. Error t value Pr(>|t|)
##
                       Estimate
## triceps 4.2637045669999400488 2.9665382148454386702 1.437 0.170
## thigh
           -2.9287006520636991169 2.6469556563935556781 -1.106 0.285
## midarm
          -1.5614167939150791486 1.1396021351584755266 -1.370 0.190
## Residual standard error: 0.4857 on 16 degrees of freedom
## Multiple R-squared: 0.8014, Adjusted R-squared: 0.7641
## F-statistic: 21.52 on 3 and 16 DF, p-value: 0.000007343
# Test for multicollinearity
library(olsrr)
##
## Attaching package: 'olsrr'
## The following object is masked from 'package:datasets':
##
##
    rivers
ols_coll_diag(bodyfat_lm_full)
## Tolerance and Variance Inflation Factor
## -----
## Variables Tolerance VIF
## 1 triceps 0.001410750 708.8429
## 2 thigh 0.001771971 564.3434
## 3 midarm 0.009559681 104.6060
##
##
## Eigenvalue and Condition Index
## -----
## Eigenvalue Condition Index intercept triceps thigh
## 2 0.020522792222
                    13.90482 0.000371520161 0.001319092929 0.000032620187
## 3 0.011511821361
                    18.56570 0.000599149515 0.000218746554 0.000325502370
                   677.37207 0.999027383645 0.998458964415 0.999640773192
## 4 0.000008647934
     midarm
## 1 0.000009798156
## 2 0.001388740360
## 3 0.006933507626
## 4 0.991667953858
```