



University of St.Gallen

Penalized Regression

University of St. Gallen
School of Management, Economics, Law,
Social Sciences, International Affairs
and Computer Science

Assignment 2

Data Analytics I: Predictive Econometrics
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submitted by

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Requirements

To solve the following tasks, the required libraries and the data sets are loaded first.

```
library(glmnet)
library(corrplot)
library(ggplot2)
library(dplyr)

load("GHA/student-mat-train.RData")
load("GHA/student-mat-test.RData")
```

Exercise 1

There are 214 observations in the training data set and 143 observations in the test data set.

```
(n_obs_train <- nrow(train))
```

```
## [1] 214
```

```
(n_obs_test <- nrow(test))
```

```
## [1] 143
```

Exercise 2

The average grade is ~11.64, the minimum grade is 4 and the maximum grade is 19. All numbers were calculated using the training data.

```
(avg_grade <- mean(train$G3))
```

```
## [1] 11.64019
```

```
(min_grade <- min(train$G3))
```

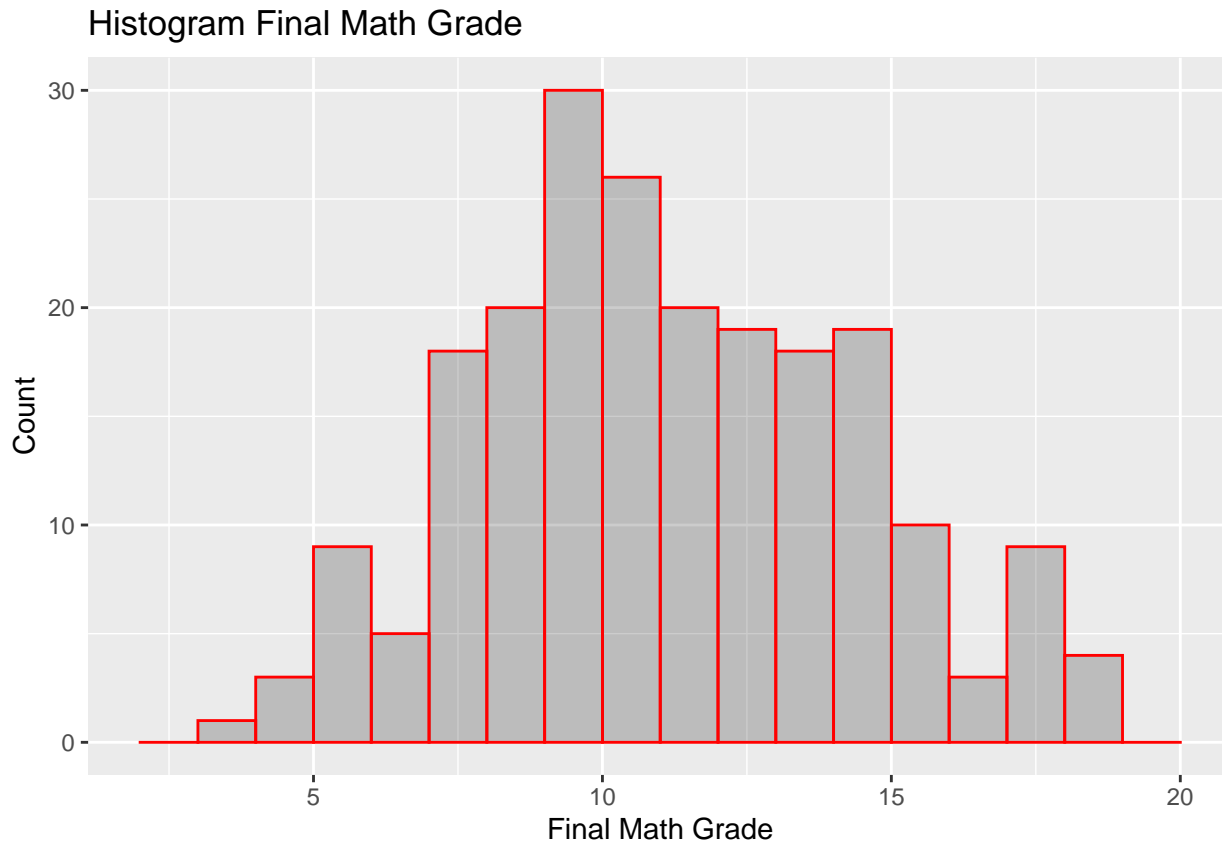
```
## [1] 4
```

```
(max_grade <- max(train$G3))
```

```
## [1] 19
```

Exercise 3

```
(final_grade_hist <- ggplot(data=train, aes(G3)) +
  geom_histogram(breaks=seq(2,20, by=1),
    col="red",
    fill="black",
    alpha = 0.2)+
  labs(title="Histogram Final Math Grade", x="Final Math Grade", y="Count"))
```



Exercise 4

When doing causal modeling there are independent variables (x_1, \dots, x_n) which are considered as the cause of the dependent variable (y), therefore one would expect a direct impact of the independent variables on the dependent variable. For predictive modelling the goal is to establish a method that allows to make predictions of the dependent variable (y) based on the known independent variables (x_1, \dots, x_n).

Exercise 5

```
(OLS1 <- lm(G3 ~ . ,
  data=select(train, G3, Medu, Fedu, studytime, schoolsup, higher)))
```

```
##
```

```
## Call:
## lm(formula = G3 ~ ., data = select(train, G3, Medu, Fedu, studytime,
##   schoolsup, higher))
##
## Coefficients:
## (Intercept)      Medu      Fedu  studytime  schoolsup    higher
##    9.38701    0.36742    0.07675    0.60662   -3.36832    0.77327
```

```
(summary(OLS1))
```

```
##
## Call:
## lm(formula = G3 ~ ., data = select(train, G3, Medu, Fedu, studytime,
##   schoolsup, higher))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -8.4668 -2.1690 -0.1981  2.0630  7.0630
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  9.38701     1.05127   8.929 2.29e-16 ***
## Medu         0.36742     0.24753   1.484  0.1392
## Fedu         0.07675     0.24727   0.310  0.7566
## studytime    0.60662     0.24803   2.446  0.0153 *
## schoolsup    -3.36832     0.67412  -4.997 1.24e-06 ***
## higher       0.77327     1.02224   0.756  0.4502
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.041 on 208 degrees of freedom
## Multiple R-squared:  0.1501, Adjusted R-squared:  0.1297
## F-statistic: 7.346 on 5 and 208 DF,  p-value: 2.312e-06
```

```
(OLS2 <- lm(G3 ~ . + .^2,
            data=select(train, G3, Medu, Fedu, studytime, schoolsup, higher)))
```

```
##
## Call:
## lm(formula = G3 ~ . + .^2, data = select(train, G3, Medu, Fedu,
##   studytime, schoolsup, higher))
##
## Coefficients:
## (Intercept)      Medu      Fedu
##    13.172132    0.145263   -1.427466
## studytime  schoolsup    higher
```

```
##          -0.464677          1.920704          -4.432522
##          Medu:Fedu          Medu:studytime          Medu:schoolsup
##          -0.001922          0.105788          -2.611720
##          Medu:higher          Fedu:studytime          Fedu:schoolsup
##          0.322940          -0.499887          1.271388
##          Fedu:higher studytime:schoolsup          studytime:higher
##          1.939871          -0.210424          2.074920
##          schoolsup:higher
##          -1.165641
```

```
(summary(OLS2))
```

```
##
## Call:
## lm(formula = G3 ~ . + .^2, data = select(train, G3, Medu, Fedu,
##      studytime, schoolsup, higher))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -8.6603 -2.0887 -0.0921  1.8277  7.8154
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    13.172132     5.437629   2.422  0.01632 *
## Medu             0.145263     1.064137   0.137  0.89156
## Fedu            -1.427466     2.050517  -0.696  0.48715
## studytime       -0.464677     2.785748  -0.167  0.86769
## schoolsup        1.920704     4.596735   0.418  0.67652
## higher          -4.432522     5.533668  -0.801  0.42409
## Medu:Fedu       -0.001922     0.217956  -0.009  0.99297
## Medu:studytime   0.105788     0.312408   0.339  0.73525
## Medu:schoolsup   -2.611720     0.899135  -2.905  0.00409 **
## Medu:higher      0.322940     1.040274   0.310  0.75656
## Fedu:studytime   -0.499887     0.298323  -1.676  0.09538 .
## Fedu:schoolsup    1.271388     0.844657   1.505  0.13386
## Fedu:higher      1.939871     2.119975   0.915  0.36128
## studytime:schoolsup -0.210424     0.851999  -0.247  0.80518
## studytime:higher  2.074920     2.739937   0.757  0.44978
## schoolsup:higher  -1.165641     4.726445  -0.247  0.80546
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 3.01 on 198 degrees of freedom
## Multiple R-squared:  0.2076, Adjusted R-squared:  0.1476
## F-statistic: 3.459 on 15 and 198 DF, p-value: 3.007e-05
```

Exercise 6

```
(OLS3 <- lm(G3 ~ . ,
            data=select(train, G3, Medu, Fedu, studytime, schoolsup, higher, Pstatus, famrel, fai

##
## Call:
## lm(formula = G3 ~ ., data = select(train, G3, Medu, Fedu, studytime,
##   schoolsup, higher, Pstatus, famrel, failures, famsup, internet))
##
## Coefficients:
## (Intercept)      Medu      Fedu  studytime  schoolsup      higher
##   9.649486    0.350090    0.007509    0.597455   -3.151785    0.284839
##   Pstatus      famrel    failures      famsup    internet
##   0.022675    0.272672   -1.016545   -0.891842    0.562597

(summary(OLS3))

##
## Call:
## lm(formula = G3 ~ ., data = select(train, G3, Medu, Fedu, studytime,
##   schoolsup, higher, Pstatus, famrel, failures, famsup, internet))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -7.0295 -2.1703 -0.0742  1.9681  7.1631
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  9.649486   1.345550   7.171 1.36e-11 ***
## Medu         0.350090   0.248736   1.407  0.1608
## Fedu         0.007509   0.242057   0.031  0.9753
## studytime    0.597455   0.247898   2.410  0.0168 *
## schoolsup    -3.151785   0.657969  -4.790 3.21e-06 ***
## higher       0.284839   1.002100   0.284  0.7765
## Pstatus      0.022675   0.610754   0.037  0.9704
## famrel       0.272672   0.230688   1.182  0.2386
## failures    -1.016545   0.317847  -3.198  0.0016 **
## famsup      -0.891842   0.449153  -1.986  0.0484 *
## internet     0.562597   0.542256   1.038  0.3007
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.943 on 203 degrees of freedom
## Multiple R-squared:  0.2233, Adjusted R-squared:  0.185
## F-statistic: 5.836 on 10 and 203 DF, p-value: 1.013e-07
```

```
(OLS4 <- lm(G3 ~ . + .^2,
            data=select(train, G3, Medu, Fedu, studytime, schoolsup, higher, Pstatus, famrel, fai
```

```
##
## Call:
## lm(formula = G3 ~ . + .^2, data = select(train, G3, Medu, Fedu,
##      studytime, schoolsup, higher, Pstatus, famrel, failures,
##      famsup, internet))
##
## Coefficients:
##      (Intercept)              Medu              Fedu
##           8.00486           2.65008          -2.33596
##      studytime      schoolsup           higher
##        -0.74497        -1.03745         0.15984
##        Pstatus          famrel          failures
##       -3.81423         1.04447         0.04181
##        famsup          internet      Medu:Fedu
##         3.05249        -1.90817        -0.03842
##      Medu:studytime  Medu:schoolsup  Medu:higher
##        -0.15811        -2.85143        -0.54863
##      Medu:Pstatus    Medu:famrel    Medu:failures
##        -0.91279        -0.36014        -1.01657
##      Medu:famsup     Medu:internet  Fedu:studytime
##         0.74973        -0.59083        -0.32178
##      Fedu:schoolsup   Fedu:higher   Fedu:Pstatus
##         1.41207         1.45072         0.10424
##      Fedu:famrel     Fedu:failures  Fedu:famsup
##         0.22857         0.34034        -0.49855
##      Fedu:internet  studytime:schoolsup  studytime:higher
##         1.11916        -0.43875         1.74282
##      studytime:Pstatus  studytime:famrel  studytime:failures
##         1.28299         0.09156        -0.61083
##      studytime:famsup  studytime:internet  schoolsup:higher
##         0.79983         0.10897         0.67617
##      schoolsup:Pstatus  schoolsup:famrel  schoolsup:failures
##        -0.95843         0.16037         1.93873
##      schoolsup:famsup  schoolsup:internet  higher:Pstatus
##         1.18057         0.27075         2.66107
##      higher:famrel    higher:failures  higher:famsup
##        -0.14551         1.51356        -4.68077
##      higher:internet  Pstatus:famrel  Pstatus:failures
##           NA           1.04521        -0.19850
##      Pstatus:famsup   Pstatus:internet  famrel:failures
##        -0.71384         0.89706        -0.26244
##      famrel:famsup    famrel:internet  failures:famsup
##        -0.72776         0.21125         0.67703
##      failures:internet  famsup:internet
```



```
##          -0.39309          0.95906
```

```
(summary(OLS4))
```

```
##
## Call:
## lm(formula = G3 ~ . + .^2, data = select(train, G3, Medu, Fedu,
##      studytime, schoolsup, higher, Pstatus, famrel, failures,
##      famsup, internet))
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -6.302 -1.638  0.000  1.569  7.129
##
## Coefficients: (1 not defined because of singularities)
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    8.00486   19.71735    0.406  0.68530
## Medu           2.65008    2.18285    1.214  0.22653
## Fedu          -2.33596    5.62321   -0.415  0.67840
## studytime     -0.74497    5.53425   -0.135  0.89309
## schoolsup     -1.03745   10.53794   -0.098  0.92170
## higher         0.15984   19.57437    0.008  0.99350
## Pstatus       -3.81423    6.70935   -0.568  0.57050
## famrel         1.04447    4.90832    0.213  0.83176
## failures       0.04181    2.33857    0.018  0.98576
## famsup         3.05249    5.01243    0.609  0.54340
## internet      -1.90817    2.90639   -0.657  0.51242
## Medu:Fedu      -0.03842    0.25236   -0.152  0.87919
## Medu:studytime -0.15811    0.34987   -0.452  0.65196
## Medu:schoolsup -2.85143    1.06173   -2.686  0.00801 **
## Medu:higher    -0.54863    1.68739   -0.325  0.74551
## Medu:Pstatus   -0.91279    0.98786   -0.924  0.35689
## Medu:famrel    -0.36014    0.40866   -0.881  0.37951
## Medu:failures  -1.01657    0.63945   -1.590  0.11388
## Medu:famsup     0.74973    0.61857    1.212  0.22730
## Medu:internet  -0.59083    0.67187   -0.879  0.38052
## Fedu:studytime -0.32178    0.33980   -0.947  0.34510
## Fedu:schoolsup  1.41207    0.96285    1.467  0.14448
## Fedu:higher     1.45072    5.53069    0.262  0.79343
## Fedu:Pstatus    0.10424    0.80597    0.129  0.89725
## Fedu:famrel     0.22857    0.38702    0.591  0.55563
## Fedu:failures   0.34034    0.67323    0.506  0.61389
## Fedu:famsup    -0.49855    0.63597   -0.784  0.43426
## Fedu:internet   1.11916    0.71180    1.572  0.11787
## studytime:schoolsup -0.43875    0.94950   -0.462  0.64465
## studytime:higher  1.74282    5.29531    0.329  0.74249
## studytime:Pstatus  1.28299    1.13311    1.132  0.25923
```

```
## studytime:famrel      0.09156      0.28279      0.324      0.74654
## studytime:failures    -0.61083      0.81245     -0.752      0.45326
## studytime:famsup      0.79983      0.63934      1.251      0.21277
## studytime:internet    0.10897      0.81962      0.133      0.89440
## schoolsup:higher      0.67617      9.41038      0.072      0.94281
## schoolsup:Pstatus     -0.95843      2.60822     -0.367      0.71376
## schoolsup:famrel      0.16037      1.43567      0.112      0.91120
## schoolsup:failures    1.93873      1.64480      1.179      0.24028
## schoolsup:famsup      1.18057      2.35659      0.501      0.61709
## schoolsup:internet    0.27075      2.11487      0.128      0.89830
## higher:Pstatus        2.66107      4.74288      0.561      0.57554
## higher:famrel         -0.14551      4.88194     -0.030      0.97626
## higher:failures       1.51356      1.98140      0.764      0.44607
## higher:famsup         -4.68077      4.62750     -1.012      0.31331
## higher:internet       NA           NA           NA           NA
## Pstatus:famrel        1.04521      0.69232      1.510      0.13310
## Pstatus:failures     -0.19850      1.71396     -0.116      0.90795
## Pstatus:famsup        -0.71384      1.84148     -0.388      0.69880
## Pstatus:internet      0.89706      1.95701      0.458      0.64730
## famrel:failures       -0.26244      0.39874     -0.658      0.51138
## famrel:famsup         -0.72776      0.67389     -1.080      0.28180
## famrel:internet       0.21125      0.72523      0.291      0.77121
## failures:famsup       0.67703      0.98499      0.687      0.49286
## failures:internet     -0.39309      1.02403     -0.384      0.70159
## famsup:internet       0.95906      1.41469      0.678      0.49880
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.919 on 159 degrees of freedom
## Multiple R-squared:  0.4014, Adjusted R-squared:  0.1981
## F-statistic: 1.974 on 54 and 159 DF,  p-value: 0.0006072
```

```
MSE_IS_OLS1 <- mean((train$G3 - OLS1$fitted.values)^2)
MSE_IS_OLS2 <- mean((train$G3 - OLS2$fitted.values)^2)
MSE_IS_OLS3 <- mean((train$G3 - OLS3$fitted.values)^2)
MSE_IS_OLS4 <- mean((train$G3 - OLS4$fitted.values)^2)
```

```
fit_OLS1 <- predict(OLS1, newdata = test)
MSE_OOS_OLS1 <- mean((test$G3 - fit_OLS1)^2)
```

```
fit_OLS2 <- predict(OLS2, newdata = test)
MSE_OOS_OLS2 <- mean((test$G3 - fit_OLS2)^2)
```

```
fit_OLS3 <- predict(OLS3, newdata = test)
MSE_OOS_OLS3 <- mean((test$G3 - fit_OLS3)^2)
```

```
fit_OLS4 <- predict(OLS4, newdata = test)
MSE_OOS_OLS4 <- mean((test$G3 - fit_OLS4)^2)
```

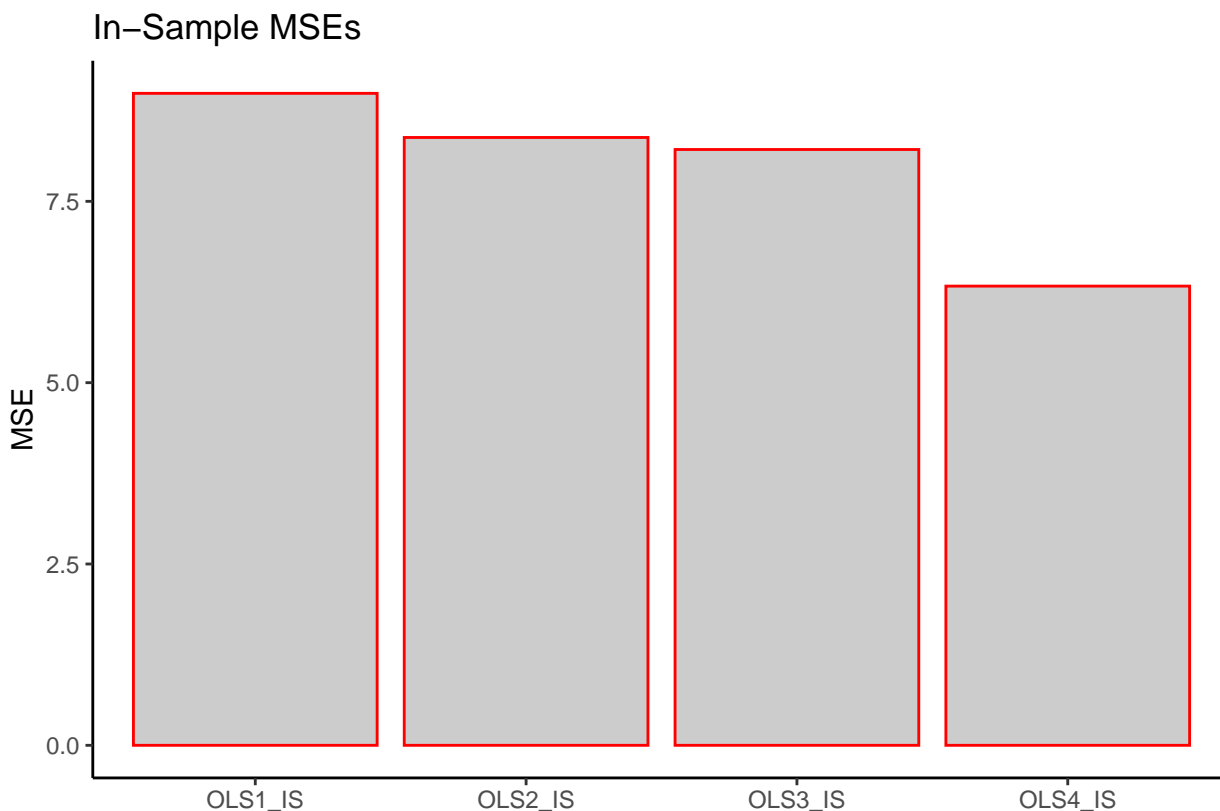
```
(MSE_IS <- data.frame(model = c("OLS1_IS", "OLS2_IS", "OLS3_IS", "OLS4_IS"),
                        MSE = c(MSE_IS_OLS1, MSE_IS_OLS2, MSE_IS_OLS3, MSE_IS_OLS4)))
```

```
##      model      MSE
## 1 OLS1_IS 8.988840
## 2 OLS2_IS 8.380403
## 3 OLS3_IS 8.214626
## 4 OLS4_IS 6.330987
```

```
(MSE_OOS <- data.frame(model = c("OLS1_OOS", "OLS2_OOS", "OLS3_OOS", "OLS4_OOS"),
                        MSE = c(MSE_OOS_OLS1, MSE_OOS_OLS2, MSE_OOS_OLS3, MSE_OOS_OLS4)))
```

```
##      model      MSE
## 1 OLS1_OOS 10.103001
## 2 OLS2_OOS 10.467642
## 3 OLS3_OOS  9.709007
## 4 OLS4_OOS 12.466627
```

```
(ggplot(MSE_IS, aes(model, MSE)) +
  geom_col(color = "red", fill = 'black', alpha = 0.2) +
  ggtitle("In-Sample MSEs") +
  xlab("") +
  theme_classic())
```



```
(ggplot(MSE_OOS, aes(model, MSE)) +  
  geom_col(color = "red", fill = 'black', alpha = 0.2) +  
  ggtitle("Out-Of-Sample MSEs") +  
  xlab("") +  
  theme_classic())
```

