HW08

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Exericse 1

-5

-3

Log Lambda

-2

-1

```
library(readr)
leukemia = read_csv("/Users/Constance/Downloads/leukemia.csv", progress = FALSE)
y = as.factor(leukemia$class)
X = as.matrix(leukemia[, -1])
library(glmnet)
## Loading required package: Matrix
## Loading required package: foreach
## Loaded glmnet 2.0-13
fit_lasso = glmnet(X, y, alpha = 1, family="binomial")
fit_ridge = glmnet(X, y, alpha = 0, family="binomial")
par(mfrow = c(1, 2))
plot(fit_lasso, xvar = "lambda", main = "Lasso")
plot(fit_ridge, xvar = "lambda", main = "Ridge")
                                                                        Ridge
                     Lasso 2
                                 13
                                                               5147
               27
                                        1
                                                                                       5147
      0.002
                                                       0.00010
Coefficients
                                                 Coefficients
      0.000
                                                       0.0000.0
      -0.002
                                                       -0.00015
                                                                 2
```

3

4

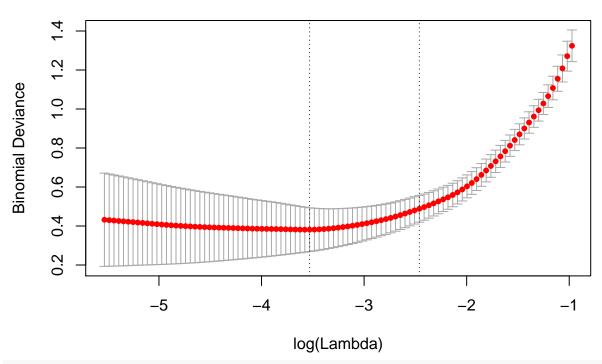
Log Lambda

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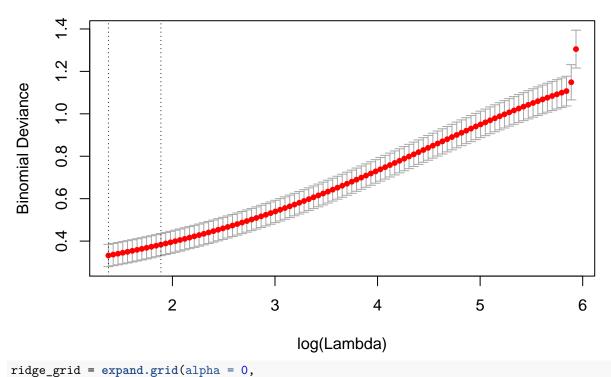
```
fit_cv = cv.glmnet(X, y, family = "binomial", alpha = 1)
plot(fit_cv)
```

28 27 27 27 25 25 23 21 19 18 10 8 5 1 0



library(caret)

5147 5147 5147 5147 5147 5147 5147 5147



```
lambda = c(fit_cv2$lambda.min, fit_cv2$lambda.1se))

set.seed(659017838)
fit_ridge = train(X, y,
    method = "glmnet",
    trControl = cv_5,
    tuneGrid = ridge_grid
)

ridge_accuracies = fit_ridge$results$Accuracy

set.seed(659017838)
fit_knn = train(X, y,
    method = "knn",
    trControl = cv_5,
    preProcess = c("center", "scale")
)
```

Model	Cross-Validated Accuracy	Standard Deviation
fit_lasso	0.9152381	0.0600831
fit_lasso	0.9009524	0.0818854
fit_ridge	0.9428571	0.0931315
fit_ridge	0.9428571	0.0931315
fit_knn	0.8885714	0.0637633
fit_knn	0.8742857	0.0935929
fit_knn	0.832381	0.0956776

knn_accuracies = fit_knn\$results\$Accuracy

Exercise 2

```
set.seed(42)
library(caret)
library(ISLR)
index = createDataPartition(College$Outstate, p = 0.75, list = FALSE)
college trn = College[index, ]
college_tst = College[-index, ]
set.seed(659017838)
linear mod = train(
 Outstate ~ .,
  data = college_trn,
 method = "lm",
 trControl = trainControl(method = "cv", number = 5)
set.seed(659017838)
elnet1 = train(
  Outstate ~ . ,
  data = college_trn,
 method = "glmnet",
 trControl = cv_5,
  tuneLength = 10
set.seed(659017838)
elnet2 = train(
  Outstate ~ .^2 ,
 data = college_trn,
 method = "glmnet",
 trControl = cv_5,
  tuneLength = 10
)
set.seed(659017838)
knn_mod = train(
  Outstate ~ .,
 data = college_trn,
 method = "knn",
 trControl = trainControl(method = "cv", number = 5),
  preProcess = c("center", "scale")
set.seed(659017838)
knn_mod2 = train(
 Outstate ~ .^2,
 data = college_trn,
 method = "knn",
 trControl = trainControl(method = "cv", number = 5),
  preProcess = c("center", "scale")
library(randomForest)
set.seed(659017838)
```

```
rf_mod = train(
  Outstate ~ .,
  data = college_trn,
  method = "rf",
  trControl = trainControl(method = "cv", number = 5)
get_best_result = function(caret_fit) {
  best = which(rownames(caret_fit$results) == rownames(caret_fit$bestTune))
  best_result = caret_fit$results[best, ]
  rownames(best_result) = NULL
  best_result
models = list(linear_mod, elnet1, elnet2, knn_mod, knn_mod2, rf_mod)
rmse = function(actual, predicted) {
  sqrt(mean((actual - predicted) ^ 2))
}
get_rmse = function(model, data, response) {
  rmse(actual = data[, response],
       predicted = predict(model, data))
test_rmse = sapply(models, get_rmse, data = college_tst, response = "Outstate")
```

Model	Cross-Validated RMSE	Test RMSE
linear_mod	1974.7423144	2069.0871599
elnet1	1970.6806276	2090.9905103
elnet2	1844.394725	1964.7392123
knn_mod	1905.530466	2007.5781377
knn_mod2	1971.5648606	2060.8032232
rf_mod	1749.1556016	1713.9861725

Exercise 3

```
uiuc_outstate = College[rownames(College)=="University of Illinois - Urbana",'Outstate']
```

Question Answer (a) 72 observations are in the dataset, and 5147 predictors are in the dataset. (b) Yes. We see a nice U-shaped CV error curve (c) No. This plot suggests that if we were to try smaller lambda, we could achieve a lower devience (error). (d) KNN performs worse. This is expected in a high-dimensional setting due to the curse of dimensionality. The model with a ridge penalty should be choicen because it has the largest accuracy. (e) (f) I prefer the random forest model because it has the smallest cross-validated RMSE and test RMSE. (g) The first elastic net model: alpha: 0.1, lambda: 82.9579489 The second elastic net model: alpha: 0.1, lambda: 220.8521303

Both have an alpha value of 0.1, so are in-between, but are closer to ridge.

Question	Answer
(h)	Yes. Yes. A lower error is found using scaled predictors.
(i)	Without interactions seems to work better. Adding all the interactions creates a high
	dimensional dataset.
(j)	The dataset is from year 1995, and the out-of-state tuition at UIUC at that time was 7560.