

# STAT 432: Basics of Statistical Learning

## *Quiz II - Review Questions*

### Exercise 1

Two models were fit using **ridge** regression. One with  $\lambda = 0.1$  and the other with  $\lambda = 0.002$ . Which was fit using  $\lambda = 0.1$ ?

Model **A**:

```
## (Intercept)      x1      x2      x3      x4      x5
## -0.7618        0.3647  0.1893  1.0853  0.7744  1.3136
```

Model **B**:

```
## (Intercept)      x1      x2      x3      x4      x5
## -0.5782        0.3322  0.1869  0.9741  0.6806  1.1916
```

### Exercise 2

Two models were fit using the **lasso**. One with  $\lambda = 0.1$  and the other with  $\lambda = 0.002$ . Which was fit using  $\lambda = 0.1$ ?

Model **A**:

```
## (Intercept)      x1      x2      x3      x4      x5
## 0.2709        0.0000  0.0000  0.5547  0.2724  0.8676
```

Model **B**:

```
## (Intercept)      x1      x2      x3      x4      x5
## -0.7445        0.3581  0.1838  1.0772  0.7665  1.3075
```

### Exercise 3

Two models were selected, one using backwards AIC and the other using backwards BIC. Which was found using backwards BIC?

Model **A**:

```
## (Intercept)      x5
## 0.5515        1.1272
```

Model **B**:

```
## (Intercept)      x3      x4      x5
## -0.4745        1.0845  0.7446  1.3418
```

## Exercise 4

Suppose we estimate the regression coefficients in a linear regression model by minimizing

$$\sum_{i=1}^n \left( y_i - \beta_0 - \sum_{j=1}^p \beta_j x_{ij} \right)^2 \text{ subject to } \sum_{j=1}^p \beta_j^2 \leq s.$$

for a particular value of  $s$ . Draw a plot that indicates how train and test RMSE are affected by the choice of  $s$ . Assume that simply minimizing RSS would result in a model that overfits. Also assume that a model with only the intercept would underfit.

## Exercise 5

Continue with the setup from Exercise 4. Consider using this model to make a prediction. What happens to **variance** as  $s$  is increased from 0?

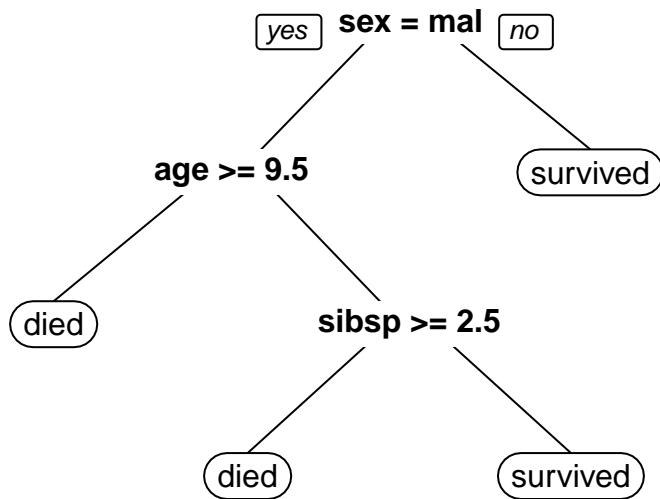
## Exercise 6

Continue with the setup from Exercise 4. Consider using this model to make a prediction. What happens to **squared bias** as  $s$  is increased from 0?

## Exercise 7

A tree model was fit to training data for a classification problem. Below is a sample of the data used, where `survived` is the response. We are attempting to predict the survival status of passengers of the Titanic:

pclass	survived	sex	age	sibsp	parch
1st	survived	female	29.0000	0	0
1st	survived	male	0.9167	1	2
1st	died	female	2.0000	1	2
1st	died	male	30.0000	1	2
1st	died	female	25.0000	1	2



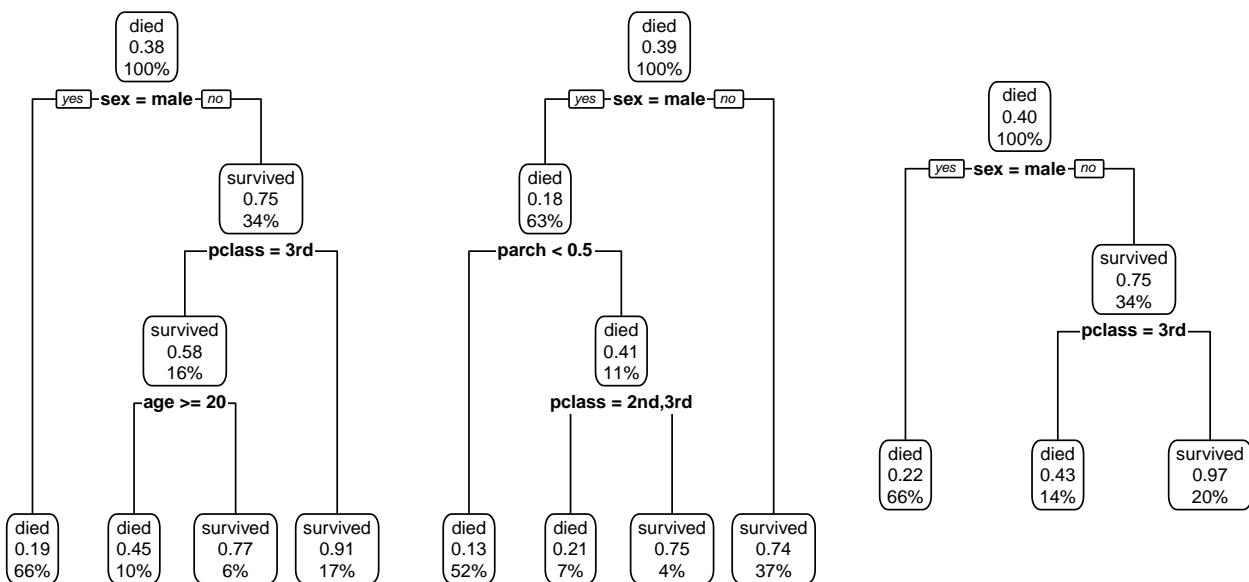
Predict the outcome of `survived` for the following three test passengers:

pclass	sex	age	sibsp	parch
1st	male	71	0	0
1st	female	48	1	0
2nd	male	3	1	1

## Exercise 8

A bagged tree model (using a total of 3 trees) was fit to training data for a classification problem. Below is a sample of the data used, where `survived` is the response. We are attempting to predict the survival status of passengers of the Titanic:

pclass	survived	sex	age	sibsp	parch
1st	survived	female	29.0000	0	0
1st	survived	male	0.9167	1	2
1st	died	female	2.0000	1	2
1st	died	male	30.0000	1	2
1st	died	female	25.0000	1	2

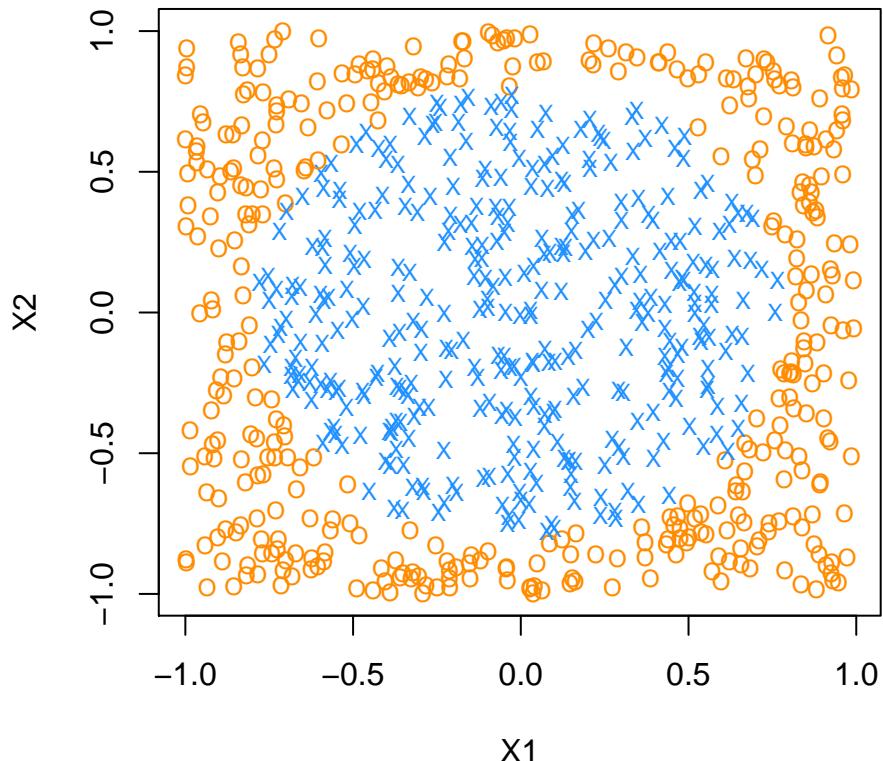


Predict the outcome of `survived`, and predicted probability of surviving for the following test passengers:

pclass	sex	age	sibsp	parch
1st	male	71	0	0
1st	female	48	1	0

### Exercise 9

Consider fitting classification methods to the data in the plot below. For each part below, specify which of the two methods will perform better.



- (a) Additive Logistic Regression or  $k$ -Nearest Neighbors?
- (b) Additive Logistic Regression or a single Decision Tree?
- (c) Additive Logistic Regression or a Random Forest?
- (d) A single decision tree or a Random Forest?

## Solutions

- 1. B
- 2. A
- 3. A
- 4. Test: U-shaped. Train: Decrease as  $s$  increases.
- 5. Increase.
- 6. Decrease.
- 7. Died. Survived. Survived.
- 8. Died. (0.18) Survived. (0.8733)
- 9. KNN. Tree. RF. RF.