

# 7390–Machine Learning and Data Sciences

Northeastern University, Fall 2017

FINAL EXAM (PRACTICE), SATURDAY DEC 09, 2017

- The exam is open book.
- You are given 60 minutes for this exam.
- **Show your work.** Partial credit will be given.

Name: \_\_\_\_\_

1. For each of parts (a) through (d), indicate whether we would generally expect the performance of a flexible statistical learning method to be better or worse than an inflexible method. Justify your answer.
  - (a) The sample size  $n$  is extremely large, and the number of predictors  $p$  is small.
  - (b) The number of predictors  $p$  is extremely large, and the number of observations  $n$  is small.
  - (c) The relationship between the predictors and response is highly non-linear.
  - (d) The variance of the error terms, i.e.  $\sigma^2 = \text{Var}(\epsilon)$ , is extremely high.
2. Explain whether each scenario is a classification or regression problem, and indicate whether we are most interested in inference or prediction. Finally, provide  $n$  (number of observations) and  $p$  (number of features).
  - (a) We collect a set of data on the top 500 firms in the US. For each firm we record profit, number of employees, industry and the CEO salary. We are interested in understanding which factors affect CEO salary.
  - (b) We are considering launching a new product and wish to know whether it will be a success or a failure. We collect data on 20 similar products that were previously launched. For each product we have recorded whether it was a success or failure, price charged for the product, marketing budget, competition price, and ten other variables.
3. What are the advantages and disadvantages of a very flexible (versus a less flexible) approach for regression or classification? Under what circumstances might a more flexible approach be preferred to a less flexible approach? When might a less flexible approach be preferred?
4. Suppose we collect data for a group of students in a statistics class with variables  $X_1$  = hours studied,  $X_2$  = undergrad GPA, and  $Y$  = receive an A. We fit a logistic regression and produce estimated coefficient,  $\beta_0 = 6$ ,  $\beta_1 = 0.05$ ,  $\beta_2 = 1$ .
  - (a) Estimate the probability that a student who studies for 40 h and has an undergrad GPA of 3.5 gets an A in the class.
  - (b) How many hours would the student in part (a) need to study to have a 50 % chance of getting an A in the class?

The estimated probability that a student gets an A in the class on a test data is given in the following table.

ID	$X_1$	$X_2$	$Y$	Estimated Probability
1	50	3.9	A	0.60
2	45	3.7	A	0.49
3	42	3.9	A	0.5
4	40	3.5	not A	0.38
5	38	3.4	not A	0.33
6	36	3.2	not A	0.27
7	35	3.8	A	0.39
8	34	3.1	not A	0.23
9	32	3.0	not A	0.20
10	31	3.1	A	0.21
11	30	2.9	not A	0.17
12	28	3.7	A	0.29
13	28	3.3	not A	0.21
14	25	2.9	not A	0.16
15	20	2.8	not A	0.10

- (c) Estimate the test AUC on this dataset.
- (d) Calculate confusion matrix, FP and TP rates, accuracy, and balanced accuracy for threshold=0.3
5. Explain how  $k$ -fold cross-validation is implemented. What are the advantages and disadvantages of  $k$ -fold cross-validation relative to the validation set approach?
6. Suppose that we use some statistical learning method to make a prediction for the response  $Y$  for a particular value of the predictor  $X$ . Carefully describe how we might estimate the standard deviation of our prediction.
7. We perform best subset, forward stepwise, and backward stepwise selection on a single data set. For each approach, we obtain  $p + 1$  models, containing  $0, 1, 2, \dots, p$  predictors. Explain your answers:
  - (a) Which of the three models with  $k$  predictors has the smallest training RSS?
  - (b) Which of the three models with  $k$  predictors has the smallest test RSS?
  - (c) True or False:
    1. The predictors in the  $k$ -variable model identified by forward stepwise are a subset of the predictors in the  $(k + 1)$ -variable model identified by forward stepwise selection.
    2. The predictors in the  $k$ -variable model identified by backward stepwise are a subset of the predictors in the  $(k + 1)$ -variable model identified by backward stepwise selection.
    3. The predictors in the  $k$ -variable model identified by backward stepwise are a subset of the predictors in the  $(k + 1)$ -variable model identified by forward stepwise selection.
    4. The predictors in the  $k$ -variable model identified by forward stepwise are a subset of the predictors in the  $(k + 1)$ -variable model identified by backward stepwise selection.
    5. The predictors in the  $k$ -variable model identified by best subset are a subset of the predictors in the  $(k + 1)$ -variable model identified by best subset selection.
8. Indicate which of (a) through (d) is correct. Justify your answer. The lasso, relative to least squares, is:
  - (a) More flexible and hence will give improved prediction accuracy when its increase in bias is less than its decrease in variance.
  - (b) More flexible and hence will give improved prediction accuracy when its increase in variance is less than its decrease in bias.
  - (c) Less flexible and hence will give improved prediction accuracy when its increase in bias is less than its decrease in variance.

- (d) Less flexible and hence will give improved prediction accuracy when its increase in variance is less than its decrease in bias.

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

$$\sum_{i=1}^n \left( y_i - \beta_0 - \sum_{j=1}^p \beta_j x_{ij} \right)^2 + \lambda \sum_{j=1}^p \beta_j^p$$

for a particular value of  $\lambda$ . For parts (a) and (b), indicate which of 1 through 5 is correct. Justify your answer.

(a) As we increase  $\lambda$  from 0, the training RSS will:

1. Increase initially, and then eventually start decreasing in an inverted U shape.
2. Decrease initially, and then eventually start increasing in a U shape.
3. Steadily increase.
4. Steadily decrease.
5. Remain constant.

(b) Repeat (a) for test RSS.

10. What are the differences between Random Forest (RF) and boosting algorithms? True or False:

- (a) RF uses the bagging method to select random subsets, and boosting uses the boosting method
- (b) RF grows trees in parallel, while boosting is sequential
- (c) RF builds multiple independent trees, while boosting builds multiple dependent trees that take into account the fit of the previous tree

11. What are advantages and disadvantages of random forest?