

```

1  setwd("C:/Users/Min Seong Kim/Dropbox/R_programming/lecture")
2
3  # College data: Demographic characteristics, tuition, and more for USA colleges.
4
5  # Private: Public/private indicator
6  # Apps: Number of applications received
7  # Accept: Number of applicants accepted
8  # Enroll: Number of new students enrolled
9  # Top10perc: New students from top 10 % of high school class
10 # Top25perc: New students from top 25 % of high school class
11 # F.Undergrad: Number of full-time undergraduates
12 # P.Undergrad: Number of part-time undergraduates
13 # Outstate: Out-of-state tuition
14 # Room.Board: Room and board costs
15 # Books: Estimated book costs
16 # Personal: Estimated personal spending
17 # PhD: Percent of faculty with Ph.D.'s
18 # Terminal: Percent of faculty with terminal degree
19 # S.F.Ratio: Student/faculty ratio
20 # perc.alumni: Percent of alumni who donate
21 # Expend: Instructional expenditure per student
22 # Grad.Rate: Graduation rate
23
24 college <- read.csv("College.csv")
25 dim(college)
26 head(college)
27 college.2 <- college[,c(-1,-7,-8,-9,-10,-11,-12,-14,-15)]
28
29 head(college.2)
30
31 reg <- lm(perc.alumni ~ Private + Top10perc, data=college.2)
32 summary(reg)
33
34 # Beyond linear models: Logistic regression modeling
35 # We can use glm() to fit a logistic regression model. The arguments are very similar
36 # to lm()
37 # The first argument is a formula, of the form Y ~ X1 + X2 + ... + Xp, where Y is the
38 # response
39 # and X1,..., Xp are predictors.
40 # We must also specify family="binomial" to get logistic regression
41
42 table(college.2$Private)
43
44 # You may want to see which variables are associated with Private.
45 par(mfrow=c(3,3), mar=c(4,4,2,0.5))
46
47 for (j in c(2:10)) {
48   plot(college.2[,1],college.2[,j], xlab="Private",
49        main=paste("Boxplot of", colnames(college.2)[j]),
50        col="lightblue")
51 }
52
53 logistic <- glm((Private=="Yes") ~ S.F.Ratio + perc.alumni, data=college.2,
54                family="binomial")
55 summary(logistic)
56
57 # Utility functions work as before
58 # To call coefficients, fitted values, residuals, summarizing, plotting, making
59 # predictions,
60 # the utility functions coef(), fitted(), residuals(), summary(), plot(), predict()
61 # work pretty much just as with lm().
62
63 coef(logistic)
64
65 p.hat <- fitted(logistic)
66 y.hat <- round(p.hat)
67 table(y.hat, y.true=college.2$Private)
68
69 # Exercise

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65
66 # You can creat a dummy variable from a continuous variable. For example,
67 college.1$dummy_pc <- as.numeric(college.1$perc.alumni > mean(college.1$perc.alumni))
68
69 # Find the variables that provide good prediction for this dummy varaible.
70
```