

MA679 Hw3

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4.6 & 4.8 & 4.9 in the next page graph!

4.10 (a)

```
library(ISLR)
summary(Weekly)
```

```
##      Year      Lag1      Lag2      Lag3
## Min.   :1990  Min.   :-18.1950  Min.   :-18.1950  Min.   :-18.1950
## 1st Qu.:1995  1st Qu.: -1.1540  1st Qu.: -1.1540  1st Qu.: -1.1580
## Median :2000  Median :  0.2410  Median :  0.2410  Median :  0.2410
## Mean   :2000  Mean    :  0.1506  Mean    :  0.1511  Mean    :  0.1472
## 3rd Qu.:2005  3rd Qu.:  1.4050  3rd Qu.:  1.4090  3rd Qu.:  1.4090
## Max.   :2010  Max.    : 12.0260  Max.    : 12.0260  Max.    : 12.0260
##      Lag4      Lag5      Volume
## Min.   :-18.1950  Min.   :-18.1950  Min.    :0.08747
## 1st Qu.: -1.1580  1st Qu.: -1.1660  1st Qu.:0.33202
## Median :  0.2380  Median :  0.2340  Median :1.00268
## Mean    :  0.1458  Mean    :  0.1399  Mean    :1.57462
## 3rd Qu.:  1.4090  3rd Qu.:  1.4050  3rd Qu.:2.05373
## Max.    : 12.0260  Max.    : 12.0260  Max.    :9.32821
##      Today      Direction
## Min.   :-18.1950  Down:484
## 1st Qu.: -1.1540  Up  :605
## Median :  0.2410
## Mean    :  0.1499
## 3rd Qu.:  1.4050
## Max.    : 12.0260
```

According to the summary, all the lags and today have the same min and max value.

(b)

```
mod1<-glm(Direction ~ Lag1 + Lag2 + Lag3 + Lag4 + Lag5 + Volume, data = Weekly, family = binomial)
summary(mod1)
```

```
##
## Call:
## glm(formula = Direction ~ Lag1 + Lag2 + Lag3 + Lag4 + Lag5 +
##      Volume, family = binomial, data = Weekly)
##
## Deviance Residuals:
```

$$4.6 \quad (a) \quad p(x) = \frac{e^{-6 + 0.05 \times 60 + 3.5}}{(1 + e^{-6 + 0.05 \times 60 + 3.5})} = 0.3775$$

$$(b) \quad p(x) = \frac{e^{-6 + 0.05 x_1 + 3.5}}{(1 + e^{-6 + 0.05 x_1 + 3.5})} = 0.5$$

$$\Rightarrow e^{-6 + 0.05 x_1 + 3.5} = 1$$

$$x_1 = \frac{2.5}{0.05} = 50$$

4.8 In knn when $k=1$, we have training error rate = 0%
 training error rate + testing error rate = $2 \times 18\%$
 \Rightarrow testing error rate = $36\% > 30\%$

We care about the error rate of test data,
 therefore, it's better to choose logistic regression

$$4.9 \quad (a) \quad \frac{p(x)}{1-p(x)} = 0.37 \Rightarrow p(x) = \frac{0.37}{1+0.37} \approx 0.27$$

$$(b) \quad \frac{p(x)}{1-p(x)} = \frac{0.16}{1-0.16} \approx 0.19$$

Figure 1: lalala

```
##      Min      1Q   Median      3Q      Max
## -1.6949 -1.2565  0.9913   1.0849   1.4579
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  0.26686    0.08593   3.106  0.0019 **
## Lag1        -0.04127    0.02641  -1.563  0.1181
## Lag2         0.05844    0.02686   2.175  0.0296 *
## Lag3        -0.01606    0.02666  -0.602  0.5469
## Lag4        -0.02779    0.02646  -1.050  0.2937
## Lag5        -0.01447    0.02638  -0.549  0.5833
## Volume      -0.02274    0.03690  -0.616  0.5377
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 1496.2  on 1088  degrees of freedom
## Residual deviance: 1486.4  on 1082  degrees of freedom
## AIC: 1500.4
##
## Number of Fisher Scoring iterations: 4
# From the model summary, we can see that only the intercept and Lag2 are significant.
```

(c)

```
library(caret)

## Loading required package: lattice
## Loading required package: ggplot2
pred<-predict(mod1, type = "response")
updown <- rep("Down", length(pred))
updown[pred > 0.5] <- "Up"
table(updown, Weekly$Direction)

##
## updown Down  Up
##   Down   54  48
##    Up   430 557
# According to the confusion matrix, the accuracy of down is 54/(54+430)
# the accuracy of up is 557/(48+557)
```

(d)

```
library(dplyr)

##
## Attaching package: 'dplyr'
```

```
## The following objects are masked from 'package:stats':
##
##   filter, lag
## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union
train <- Weekly%>% filter(Year < 2009)
Weekly.20092010 <- Weekly[!train, ]

## Warning in Ops.factor(left): '!' not meaningful for factors
mod2 <- glm(Direction ~ Lag2, data = train, family = binomial)
summary(mod2)

##
## Call:
## glm(formula = Direction ~ Lag2, family = binomial, data = train)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -1.536  -1.264   1.021   1.091   1.368
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  0.20326    0.06428   3.162  0.00157 **
## Lag2         0.05810    0.02870   2.024  0.04298 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 1354.7  on 984  degrees of freedom
## Residual deviance: 1350.5  on 983  degrees of freedom
## AIC: 1354.5
##
## Number of Fisher Scoring iterations: 4
pred2<-predict(mod2, Weekly.20092010,type = "response")
updown <- rep("Down", length(pred2))
updown[pred2 > 0.5] <- "Up"
table(updown, Weekly.20092010$Direction)

##
## updown Down Up
##   Down    0  0
```

(e)

```
library(MASS)

##
## Attaching package: 'MASS'
```

```
## The following object is masked from 'package:dplyr':
##
##      select
train <- Weekly%>% filter(Year < 2009)
Weekly.20092010 <- Weekly[!train, ]
mod3 <- lda(Direction ~ Lag2, data = train, family = binomial)
mod3

## Call:
## lda(Direction ~ Lag2, data = train, family = binomial)
##
## Prior probabilities of groups:
##      Down      Up
## 0.4477157 0.5522843
##
## Group means:
##      Lag2
## Down -0.03568254
## Up    0.26036581
##
## Coefficients of linear discriminants:
##      LD1
## Lag2 0.4414162

pred3<-predict(mod3, Weekly.20092010)
table(pred3$class, Weekly.20092010$Direction)

##
##      Down Up
## Down    0  0
## Up      0  0
```

(f)

```
library(MASS)
train <- Weekly%>% filter(Year < 2009)
Weekly.20092010 <- Weekly[!train, ]
mod3 <- qda(Direction ~ Lag2, data = train, family = binomial)
mod3

## Call:
## qda(Direction ~ Lag2, data = train, family = binomial)
##
## Prior probabilities of groups:
##      Down      Up
## 0.4477157 0.5522843
##
## Group means:
##      Lag2
## Down -0.03568254
## Up    0.26036581
```

```
pred3<-predict(mod3, Weekly.20092010)
table(pred3$class, Weekly.20092010$Direction)
```

```
##
##           Down Up
## Down      0  0
## Up        0  0
```

(g)

```
library(class)
train.X <- na.omit(as.matrix(Weekly%>% filter(Year < 2009)))
test.X <- na.omit(as.matrix(Weekly%>% filter(Year >= 2009)))
train<-Weekly%>% filter(Year < 2009)
train.Direction <-na.omit(train$Direction)
set.seed(1)
###pred.knn <- knn(train.X, test.X, train.Direction, k = 1)
###table(pred.knn, Weekly.20092010$Direction)
```

4.11 (a)

```
data(Auto)
mpg<-Auto$mpg
mpg
```

```
## [1] 18.0 15.0 18.0 16.0 17.0 15.0 14.0 14.0 14.0 15.0 15.0 14.0 15.0 14.0
## [15] 24.0 22.0 18.0 21.0 27.0 26.0 25.0 24.0 25.0 26.0 21.0 10.0 10.0 11.0
## [29] 9.0 27.0 28.0 25.0 19.0 16.0 17.0 19.0 18.0 14.0 14.0 14.0 14.0 12.0
## [43] 13.0 13.0 18.0 22.0 19.0 18.0 23.0 28.0 30.0 30.0 31.0 35.0 27.0 26.0
## [57] 24.0 25.0 23.0 20.0 21.0 13.0 14.0 15.0 14.0 17.0 11.0 13.0 12.0 13.0
## [71] 19.0 15.0 13.0 13.0 14.0 18.0 22.0 21.0 26.0 22.0 28.0 23.0 28.0 27.0
## [85] 13.0 14.0 13.0 14.0 15.0 12.0 13.0 13.0 14.0 13.0 12.0 13.0 18.0 16.0
## [99] 18.0 18.0 23.0 26.0 11.0 12.0 13.0 12.0 18.0 20.0 21.0 22.0 18.0 19.0
## [113] 21.0 26.0 15.0 16.0 29.0 24.0 20.0 19.0 15.0 24.0 20.0 11.0 20.0 19.0
## [127] 15.0 31.0 26.0 32.0 25.0 16.0 16.0 18.0 16.0 13.0 14.0 14.0 14.0 29.0
## [141] 26.0 26.0 31.0 32.0 28.0 24.0 26.0 24.0 26.0 31.0 19.0 18.0 15.0 15.0
## [155] 16.0 15.0 16.0 14.0 17.0 16.0 15.0 18.0 21.0 20.0 13.0 29.0 23.0 20.0
## [169] 23.0 24.0 25.0 24.0 18.0 29.0 19.0 23.0 23.0 22.0 25.0 33.0 28.0 25.0
## [183] 25.0 26.0 27.0 17.5 16.0 15.5 14.5 22.0 22.0 24.0 22.5 29.0 24.5 29.0
## [197] 33.0 20.0 18.0 18.5 17.5 29.5 32.0 28.0 26.5 20.0 13.0 19.0 19.0 16.5
## [211] 16.5 13.0 13.0 13.0 31.5 30.0 36.0 25.5 33.5 17.5 17.0 15.5 15.0 17.5
## [225] 20.5 19.0 18.5 16.0 15.5 15.5 16.0 29.0 24.5 26.0 25.5 30.5 33.5 30.0
## [239] 30.5 22.0 21.5 21.5 43.1 36.1 32.8 39.4 36.1 19.9 19.4 20.2 19.2 20.5
## [253] 20.2 25.1 20.5 19.4 20.6 20.8 18.6 18.1 19.2 17.7 18.1 17.5 30.0 27.5
## [267] 27.2 30.9 21.1 23.2 23.8 23.9 20.3 17.0 21.6 16.2 31.5 29.5 21.5 19.8
## [281] 22.3 20.2 20.6 17.0 17.6 16.5 18.2 16.9 15.5 19.2 18.5 31.9 34.1 35.7
## [295] 27.4 25.4 23.0 27.2 23.9 34.2 34.5 31.8 37.3 28.4 28.8 26.8 33.5 41.5
## [309] 38.1 32.1 37.2 28.0 26.4 24.3 19.1 34.3 29.8 31.3 37.0 32.2 46.6 27.9
## [323] 40.8 44.3 43.4 36.4 30.0 44.6 33.8 29.8 32.7 23.7 35.0 32.4 27.2 26.6
## [337] 25.8 23.5 30.0 39.1 39.0 35.1 32.3 37.0 37.7 34.1 34.7 34.4 29.9 33.0
```



```
## [351] 33.7 32.4 32.9 31.6 28.1 30.7 25.4 24.2 22.4 26.6 20.2 17.6 28.0 27.0
## [365] 34.0 31.0 29.0 27.0 24.0 36.0 37.0 31.0 38.0 36.0 36.0 36.0 34.0 38.0
## [379] 32.0 38.0 25.0 38.0 26.0 22.0 32.0 36.0 27.0 27.0 44.0 32.0 28.0 31.0
```

```
mpg01 <- rep(0, length(mpg))
mpg01[mpg > median(mpg)] <- 1
mpg01<-data.frame(Auto,mpg01)
mpg01
```

##	mpg	cylinders	displacement	horsepower	weight	acceleration	year	origin
## 1	18.0	8	307.0	130	3504	12.0	70	1
## 2	15.0	8	350.0	165	3693	11.5	70	1
## 3	18.0	8	318.0	150	3436	11.0	70	1
## 4	16.0	8	304.0	150	3433	12.0	70	1
## 5	17.0	8	302.0	140	3449	10.5	70	1
## 6	15.0	8	429.0	198	4341	10.0	70	1
## 7	14.0	8	454.0	220	4354	9.0	70	1
## 8	14.0	8	440.0	215	4312	8.5	70	1
## 9	14.0	8	455.0	225	4425	10.0	70	1
## 10	15.0	8	390.0	190	3850	8.5	70	1
## 11	15.0	8	383.0	170	3563	10.0	70	1
## 12	14.0	8	340.0	160	3609	8.0	70	1
## 13	15.0	8	400.0	150	3761	9.5	70	1
## 14	14.0	8	455.0	225	3086	10.0	70	1
## 15	24.0	4	113.0	95	2372	15.0	70	3
## 16	22.0	6	198.0	95	2833	15.5	70	1
## 17	18.0	6	199.0	97	2774	15.5	70	1
## 18	21.0	6	200.0	85	2587	16.0	70	1
## 19	27.0	4	97.0	88	2130	14.5	70	3
## 20	26.0	4	97.0	46	1835	20.5	70	2
## 21	25.0	4	110.0	87	2672	17.5	70	2
## 22	24.0	4	107.0	90	2430	14.5	70	2
## 23	25.0	4	104.0	95	2375	17.5	70	2
## 24	26.0	4	121.0	113	2234	12.5	70	2
## 25	21.0	6	199.0	90	2648	15.0	70	1
## 26	10.0	8	360.0	215	4615	14.0	70	1
## 27	10.0	8	307.0	200	4376	15.0	70	1
## 28	11.0	8	318.0	210	4382	13.5	70	1
## 29	9.0	8	304.0	193	4732	18.5	70	1
## 30	27.0	4	97.0	88	2130	14.5	71	3
## 31	28.0	4	140.0	90	2264	15.5	71	1
## 32	25.0	4	113.0	95	2228	14.0	71	3
## 34	19.0	6	232.0	100	2634	13.0	71	1
## 35	16.0	6	225.0	105	3439	15.5	71	1
## 36	17.0	6	250.0	100	3329	15.5	71	1
## 37	19.0	6	250.0	88	3302	15.5	71	1
## 38	18.0	6	232.0	100	3288	15.5	71	1
## 39	14.0	8	350.0	165	4209	12.0	71	1
## 40	14.0	8	400.0	175	4464	11.5	71	1
## 41	14.0	8	351.0	153	4154	13.5	71	1
## 42	14.0	8	318.0	150	4096	13.0	71	1
## 43	12.0	8	383.0	180	4955	11.5	71	1
## 44	13.0	8	400.0	170	4746	12.0	71	1
## 45	13.0	8	400.0	175	5140	12.0	71	1
## 46	18.0	6	258.0	110	2962	13.5	71	1

## 47	22.0	4	140.0	72	2408	19.0	71	1
## 48	19.0	6	250.0	100	3282	15.0	71	1
## 49	18.0	6	250.0	88	3139	14.5	71	1
## 50	23.0	4	122.0	86	2220	14.0	71	1
## 51	28.0	4	116.0	90	2123	14.0	71	2
## 52	30.0	4	79.0	70	2074	19.5	71	2
## 53	30.0	4	88.0	76	2065	14.5	71	2
## 54	31.0	4	71.0	65	1773	19.0	71	3
## 55	35.0	4	72.0	69	1613	18.0	71	3
## 56	27.0	4	97.0	60	1834	19.0	71	2
## 57	26.0	4	91.0	70	1955	20.5	71	1
## 58	24.0	4	113.0	95	2278	15.5	72	3
## 59	25.0	4	97.5	80	2126	17.0	72	1
## 60	23.0	4	97.0	54	2254	23.5	72	2
## 61	20.0	4	140.0	90	2408	19.5	72	1
## 62	21.0	4	122.0	86	2226	16.5	72	1
## 63	13.0	8	350.0	165	4274	12.0	72	1
## 64	14.0	8	400.0	175	4385	12.0	72	1
## 65	15.0	8	318.0	150	4135	13.5	72	1
## 66	14.0	8	351.0	153	4129	13.0	72	1
## 67	17.0	8	304.0	150	3672	11.5	72	1
## 68	11.0	8	429.0	208	4633	11.0	72	1
## 69	13.0	8	350.0	155	4502	13.5	72	1
## 70	12.0	8	350.0	160	4456	13.5	72	1
## 71	13.0	8	400.0	190	4422	12.5	72	1
## 72	19.0	3	70.0	97	2330	13.5	72	3
## 73	15.0	8	304.0	150	3892	12.5	72	1
## 74	13.0	8	307.0	130	4098	14.0	72	1
## 75	13.0	8	302.0	140	4294	16.0	72	1
## 76	14.0	8	318.0	150	4077	14.0	72	1
## 77	18.0	4	121.0	112	2933	14.5	72	2
## 78	22.0	4	121.0	76	2511	18.0	72	2
## 79	21.0	4	120.0	87	2979	19.5	72	2
## 80	26.0	4	96.0	69	2189	18.0	72	2
## 81	22.0	4	122.0	86	2395	16.0	72	1
## 82	28.0	4	97.0	92	2288	17.0	72	3
## 83	23.0	4	120.0	97	2506	14.5	72	3
## 84	28.0	4	98.0	80	2164	15.0	72	1
## 85	27.0	4	97.0	88	2100	16.5	72	3
## 86	13.0	8	350.0	175	4100	13.0	73	1
## 87	14.0	8	304.0	150	3672	11.5	73	1
## 88	13.0	8	350.0	145	3988	13.0	73	1
## 89	14.0	8	302.0	137	4042	14.5	73	1
## 90	15.0	8	318.0	150	3777	12.5	73	1
## 91	12.0	8	429.0	198	4952	11.5	73	1
## 92	13.0	8	400.0	150	4464	12.0	73	1
## 93	13.0	8	351.0	158	4363	13.0	73	1
## 94	14.0	8	318.0	150	4237	14.5	73	1
## 95	13.0	8	440.0	215	4735	11.0	73	1
## 96	12.0	8	455.0	225	4951	11.0	73	1
## 97	13.0	8	360.0	175	3821	11.0	73	1
## 98	18.0	6	225.0	105	3121	16.5	73	1
## 99	16.0	6	250.0	100	3278	18.0	73	1
## 100	18.0	6	232.0	100	2945	16.0	73	1

## 101	18.0	6	250.0	88	3021	16.5	73	1
## 102	23.0	6	198.0	95	2904	16.0	73	1
## 103	26.0	4	97.0	46	1950	21.0	73	2
## 104	11.0	8	400.0	150	4997	14.0	73	1
## 105	12.0	8	400.0	167	4906	12.5	73	1
## 106	13.0	8	360.0	170	4654	13.0	73	1
## 107	12.0	8	350.0	180	4499	12.5	73	1
## 108	18.0	6	232.0	100	2789	15.0	73	1
## 109	20.0	4	97.0	88	2279	19.0	73	3
## 110	21.0	4	140.0	72	2401	19.5	73	1
## 111	22.0	4	108.0	94	2379	16.5	73	3
## 112	18.0	3	70.0	90	2124	13.5	73	3
## 113	19.0	4	122.0	85	2310	18.5	73	1
## 114	21.0	6	155.0	107	2472	14.0	73	1
## 115	26.0	4	98.0	90	2265	15.5	73	2
## 116	15.0	8	350.0	145	4082	13.0	73	1
## 117	16.0	8	400.0	230	4278	9.5	73	1
## 118	29.0	4	68.0	49	1867	19.5	73	2
## 119	24.0	4	116.0	75	2158	15.5	73	2
## 120	20.0	4	114.0	91	2582	14.0	73	2
## 121	19.0	4	121.0	112	2868	15.5	73	2
## 122	15.0	8	318.0	150	3399	11.0	73	1
## 123	24.0	4	121.0	110	2660	14.0	73	2
## 124	20.0	6	156.0	122	2807	13.5	73	3
## 125	11.0	8	350.0	180	3664	11.0	73	1
## 126	20.0	6	198.0	95	3102	16.5	74	1
## 128	19.0	6	232.0	100	2901	16.0	74	1
## 129	15.0	6	250.0	100	3336	17.0	74	1
## 130	31.0	4	79.0	67	1950	19.0	74	3
## 131	26.0	4	122.0	80	2451	16.5	74	1
## 132	32.0	4	71.0	65	1836	21.0	74	3
## 133	25.0	4	140.0	75	2542	17.0	74	1
## 134	16.0	6	250.0	100	3781	17.0	74	1
## 135	16.0	6	258.0	110	3632	18.0	74	1
## 136	18.0	6	225.0	105	3613	16.5	74	1
## 137	16.0	8	302.0	140	4141	14.0	74	1
## 138	13.0	8	350.0	150	4699	14.5	74	1
## 139	14.0	8	318.0	150	4457	13.5	74	1
## 140	14.0	8	302.0	140	4638	16.0	74	1
## 141	14.0	8	304.0	150	4257	15.5	74	1
## 142	29.0	4	98.0	83	2219	16.5	74	2
## 143	26.0	4	79.0	67	1963	15.5	74	2
## 144	26.0	4	97.0	78	2300	14.5	74	2
## 145	31.0	4	76.0	52	1649	16.5	74	3
## 146	32.0	4	83.0	61	2003	19.0	74	3
## 147	28.0	4	90.0	75	2125	14.5	74	1
## 148	24.0	4	90.0	75	2108	15.5	74	2
## 149	26.0	4	116.0	75	2246	14.0	74	2
## 150	24.0	4	120.0	97	2489	15.0	74	3
## 151	26.0	4	108.0	93	2391	15.5	74	3
## 152	31.0	4	79.0	67	2000	16.0	74	2
## 153	19.0	6	225.0	95	3264	16.0	75	1
## 154	18.0	6	250.0	105	3459	16.0	75	1
## 155	15.0	6	250.0	72	3432	21.0	75	1

## 156 15.0	6	250.0	72	3158	19.5	75	1
## 157 16.0	8	400.0	170	4668	11.5	75	1
## 158 15.0	8	350.0	145	4440	14.0	75	1
## 159 16.0	8	318.0	150	4498	14.5	75	1
## 160 14.0	8	351.0	148	4657	13.5	75	1
## 161 17.0	6	231.0	110	3907	21.0	75	1
## 162 16.0	6	250.0	105	3897	18.5	75	1
## 163 15.0	6	258.0	110	3730	19.0	75	1
## 164 18.0	6	225.0	95	3785	19.0	75	1
## 165 21.0	6	231.0	110	3039	15.0	75	1
## 166 20.0	8	262.0	110	3221	13.5	75	1
## 167 13.0	8	302.0	129	3169	12.0	75	1
## 168 29.0	4	97.0	75	2171	16.0	75	3
## 169 23.0	4	140.0	83	2639	17.0	75	1
## 170 20.0	6	232.0	100	2914	16.0	75	1
## 171 23.0	4	140.0	78	2592	18.5	75	1
## 172 24.0	4	134.0	96	2702	13.5	75	3
## 173 25.0	4	90.0	71	2223	16.5	75	2
## 174 24.0	4	119.0	97	2545	17.0	75	3
## 175 18.0	6	171.0	97	2984	14.5	75	1
## 176 29.0	4	90.0	70	1937	14.0	75	2
## 177 19.0	6	232.0	90	3211	17.0	75	1
## 178 23.0	4	115.0	95	2694	15.0	75	2
## 179 23.0	4	120.0	88	2957	17.0	75	2
## 180 22.0	4	121.0	98	2945	14.5	75	2
## 181 25.0	4	121.0	115	2671	13.5	75	2
## 182 33.0	4	91.0	53	1795	17.5	75	3
## 183 28.0	4	107.0	86	2464	15.5	76	2
## 184 25.0	4	116.0	81	2220	16.9	76	2
## 185 25.0	4	140.0	92	2572	14.9	76	1
## 186 26.0	4	98.0	79	2255	17.7	76	1
## 187 27.0	4	101.0	83	2202	15.3	76	2
## 188 17.5	8	305.0	140	4215	13.0	76	1
## 189 16.0	8	318.0	150	4190	13.0	76	1
## 190 15.5	8	304.0	120	3962	13.9	76	1
## 191 14.5	8	351.0	152	4215	12.8	76	1
## 192 22.0	6	225.0	100	3233	15.4	76	1
## 193 22.0	6	250.0	105	3353	14.5	76	1
## 194 24.0	6	200.0	81	3012	17.6	76	1
## 195 22.5	6	232.0	90	3085	17.6	76	1
## 196 29.0	4	85.0	52	2035	22.2	76	1
## 197 24.5	4	98.0	60	2164	22.1	76	1
## 198 29.0	4	90.0	70	1937	14.2	76	2
## 199 33.0	4	91.0	53	1795	17.4	76	3
## 200 20.0	6	225.0	100	3651	17.7	76	1
## 201 18.0	6	250.0	78	3574	21.0	76	1
## 202 18.5	6	250.0	110	3645	16.2	76	1
## 203 17.5	6	258.0	95	3193	17.8	76	1
## 204 29.5	4	97.0	71	1825	12.2	76	2
## 205 32.0	4	85.0	70	1990	17.0	76	3
## 206 28.0	4	97.0	75	2155	16.4	76	3
## 207 26.5	4	140.0	72	2565	13.6	76	1
## 208 20.0	4	130.0	102	3150	15.7	76	2
## 209 13.0	8	318.0	150	3940	13.2	76	1

## 210	19.0	4	120.0	88	3270	21.9	76	2
## 211	19.0	6	156.0	108	2930	15.5	76	3
## 212	16.5	6	168.0	120	3820	16.7	76	2
## 213	16.5	8	350.0	180	4380	12.1	76	1
## 214	13.0	8	350.0	145	4055	12.0	76	1
## 215	13.0	8	302.0	130	3870	15.0	76	1
## 216	13.0	8	318.0	150	3755	14.0	76	1
## 217	31.5	4	98.0	68	2045	18.5	77	3
## 218	30.0	4	111.0	80	2155	14.8	77	1
## 219	36.0	4	79.0	58	1825	18.6	77	2
## 220	25.5	4	122.0	96	2300	15.5	77	1
## 221	33.5	4	85.0	70	1945	16.8	77	3
## 222	17.5	8	305.0	145	3880	12.5	77	1
## 223	17.0	8	260.0	110	4060	19.0	77	1
## 224	15.5	8	318.0	145	4140	13.7	77	1
## 225	15.0	8	302.0	130	4295	14.9	77	1
## 226	17.5	6	250.0	110	3520	16.4	77	1
## 227	20.5	6	231.0	105	3425	16.9	77	1
## 228	19.0	6	225.0	100	3630	17.7	77	1
## 229	18.5	6	250.0	98	3525	19.0	77	1
## 230	16.0	8	400.0	180	4220	11.1	77	1
## 231	15.5	8	350.0	170	4165	11.4	77	1
## 232	15.5	8	400.0	190	4325	12.2	77	1
## 233	16.0	8	351.0	149	4335	14.5	77	1
## 234	29.0	4	97.0	78	1940	14.5	77	2
## 235	24.5	4	151.0	88	2740	16.0	77	1
## 236	26.0	4	97.0	75	2265	18.2	77	3
## 237	25.5	4	140.0	89	2755	15.8	77	1
## 238	30.5	4	98.0	63	2051	17.0	77	1
## 239	33.5	4	98.0	83	2075	15.9	77	1
## 240	30.0	4	97.0	67	1985	16.4	77	3
## 241	30.5	4	97.0	78	2190	14.1	77	2
## 242	22.0	6	146.0	97	2815	14.5	77	3
## 243	21.5	4	121.0	110	2600	12.8	77	2
## 244	21.5	3	80.0	110	2720	13.5	77	3
## 245	43.1	4	90.0	48	1985	21.5	78	2
## 246	36.1	4	98.0	66	1800	14.4	78	1
## 247	32.8	4	78.0	52	1985	19.4	78	3
## 248	39.4	4	85.0	70	2070	18.6	78	3
## 249	36.1	4	91.0	60	1800	16.4	78	3
## 250	19.9	8	260.0	110	3365	15.5	78	1
## 251	19.4	8	318.0	140	3735	13.2	78	1
## 252	20.2	8	302.0	139	3570	12.8	78	1
## 253	19.2	6	231.0	105	3535	19.2	78	1
## 254	20.5	6	200.0	95	3155	18.2	78	1
## 255	20.2	6	200.0	85	2965	15.8	78	1
## 256	25.1	4	140.0	88	2720	15.4	78	1
## 257	20.5	6	225.0	100	3430	17.2	78	1
## 258	19.4	6	232.0	90	3210	17.2	78	1
## 259	20.6	6	231.0	105	3380	15.8	78	1
## 260	20.8	6	200.0	85	3070	16.7	78	1
## 261	18.6	6	225.0	110	3620	18.7	78	1
## 262	18.1	6	258.0	120	3410	15.1	78	1
## 263	19.2	8	305.0	145	3425	13.2	78	1

## 264 17.7	6	231.0	165	3445	13.4	78	1
## 265 18.1	8	302.0	139	3205	11.2	78	1
## 266 17.5	8	318.0	140	4080	13.7	78	1
## 267 30.0	4	98.0	68	2155	16.5	78	1
## 268 27.5	4	134.0	95	2560	14.2	78	3
## 269 27.2	4	119.0	97	2300	14.7	78	3
## 270 30.9	4	105.0	75	2230	14.5	78	1
## 271 21.1	4	134.0	95	2515	14.8	78	3
## 272 23.2	4	156.0	105	2745	16.7	78	1
## 273 23.8	4	151.0	85	2855	17.6	78	1
## 274 23.9	4	119.0	97	2405	14.9	78	3
## 275 20.3	5	131.0	103	2830	15.9	78	2
## 276 17.0	6	163.0	125	3140	13.6	78	2
## 277 21.6	4	121.0	115	2795	15.7	78	2
## 278 16.2	6	163.0	133	3410	15.8	78	2
## 279 31.5	4	89.0	71	1990	14.9	78	2
## 280 29.5	4	98.0	68	2135	16.6	78	3
## 281 21.5	6	231.0	115	3245	15.4	79	1
## 282 19.8	6	200.0	85	2990	18.2	79	1
## 283 22.3	4	140.0	88	2890	17.3	79	1
## 284 20.2	6	232.0	90	3265	18.2	79	1
## 285 20.6	6	225.0	110	3360	16.6	79	1
## 286 17.0	8	305.0	130	3840	15.4	79	1
## 287 17.6	8	302.0	129	3725	13.4	79	1
## 288 16.5	8	351.0	138	3955	13.2	79	1
## 289 18.2	8	318.0	135	3830	15.2	79	1
## 290 16.9	8	350.0	155	4360	14.9	79	1
## 291 15.5	8	351.0	142	4054	14.3	79	1
## 292 19.2	8	267.0	125	3605	15.0	79	1
## 293 18.5	8	360.0	150	3940	13.0	79	1
## 294 31.9	4	89.0	71	1925	14.0	79	2
## 295 34.1	4	86.0	65	1975	15.2	79	3
## 296 35.7	4	98.0	80	1915	14.4	79	1
## 297 27.4	4	121.0	80	2670	15.0	79	1
## 298 25.4	5	183.0	77	3530	20.1	79	2
## 299 23.0	8	350.0	125	3900	17.4	79	1
## 300 27.2	4	141.0	71	3190	24.8	79	2
## 301 23.9	8	260.0	90	3420	22.2	79	1
## 302 34.2	4	105.0	70	2200	13.2	79	1
## 303 34.5	4	105.0	70	2150	14.9	79	1
## 304 31.8	4	85.0	65	2020	19.2	79	3
## 305 37.3	4	91.0	69	2130	14.7	79	2
## 306 28.4	4	151.0	90	2670	16.0	79	1
## 307 28.8	6	173.0	115	2595	11.3	79	1
## 308 26.8	6	173.0	115	2700	12.9	79	1
## 309 33.5	4	151.0	90	2556	13.2	79	1
## 310 41.5	4	98.0	76	2144	14.7	80	2
## 311 38.1	4	89.0	60	1968	18.8	80	3
## 312 32.1	4	98.0	70	2120	15.5	80	1
## 313 37.2	4	86.0	65	2019	16.4	80	3
## 314 28.0	4	151.0	90	2678	16.5	80	1
## 315 26.4	4	140.0	88	2870	18.1	80	1
## 316 24.3	4	151.0	90	3003	20.1	80	1
## 317 19.1	6	225.0	90	3381	18.7	80	1

## 318 34.3	4	97.0	78	2188	15.8	80	2
## 319 29.8	4	134.0	90	2711	15.5	80	3
## 320 31.3	4	120.0	75	2542	17.5	80	3
## 321 37.0	4	119.0	92	2434	15.0	80	3
## 322 32.2	4	108.0	75	2265	15.2	80	3
## 323 46.6	4	86.0	65	2110	17.9	80	3
## 324 27.9	4	156.0	105	2800	14.4	80	1
## 325 40.8	4	85.0	65	2110	19.2	80	3
## 326 44.3	4	90.0	48	2085	21.7	80	2
## 327 43.4	4	90.0	48	2335	23.7	80	2
## 328 36.4	5	121.0	67	2950	19.9	80	2
## 329 30.0	4	146.0	67	3250	21.8	80	2
## 330 44.6	4	91.0	67	1850	13.8	80	3
## 332 33.8	4	97.0	67	2145	18.0	80	3
## 333 29.8	4	89.0	62	1845	15.3	80	2
## 334 32.7	6	168.0	132	2910	11.4	80	3
## 335 23.7	3	70.0	100	2420	12.5	80	3
## 336 35.0	4	122.0	88	2500	15.1	80	2
## 338 32.4	4	107.0	72	2290	17.0	80	3
## 339 27.2	4	135.0	84	2490	15.7	81	1
## 340 26.6	4	151.0	84	2635	16.4	81	1
## 341 25.8	4	156.0	92	2620	14.4	81	1
## 342 23.5	6	173.0	110	2725	12.6	81	1
## 343 30.0	4	135.0	84	2385	12.9	81	1
## 344 39.1	4	79.0	58	1755	16.9	81	3
## 345 39.0	4	86.0	64	1875	16.4	81	1
## 346 35.1	4	81.0	60	1760	16.1	81	3
## 347 32.3	4	97.0	67	2065	17.8	81	3
## 348 37.0	4	85.0	65	1975	19.4	81	3
## 349 37.7	4	89.0	62	2050	17.3	81	3
## 350 34.1	4	91.0	68	1985	16.0	81	3
## 351 34.7	4	105.0	63	2215	14.9	81	1
## 352 34.4	4	98.0	65	2045	16.2	81	1
## 353 29.9	4	98.0	65	2380	20.7	81	1
## 354 33.0	4	105.0	74	2190	14.2	81	2
## 356 33.7	4	107.0	75	2210	14.4	81	3
## 357 32.4	4	108.0	75	2350	16.8	81	3
## 358 32.9	4	119.0	100	2615	14.8	81	3
## 359 31.6	4	120.0	74	2635	18.3	81	3
## 360 28.1	4	141.0	80	3230	20.4	81	2
## 361 30.7	6	145.0	76	3160	19.6	81	2
## 362 25.4	6	168.0	116	2900	12.6	81	3
## 363 24.2	6	146.0	120	2930	13.8	81	3
## 364 22.4	6	231.0	110	3415	15.8	81	1
## 365 26.6	8	350.0	105	3725	19.0	81	1
## 366 20.2	6	200.0	88	3060	17.1	81	1
## 367 17.6	6	225.0	85	3465	16.6	81	1
## 368 28.0	4	112.0	88	2605	19.6	82	1
## 369 27.0	4	112.0	88	2640	18.6	82	1
## 370 34.0	4	112.0	88	2395	18.0	82	1
## 371 31.0	4	112.0	85	2575	16.2	82	1
## 372 29.0	4	135.0	84	2525	16.0	82	1
## 373 27.0	4	151.0	90	2735	18.0	82	1
## 374 24.0	4	140.0	92	2865	16.4	82	1

## 375	36.0	4	105.0	74	1980	15.3	82	2
## 376	37.0	4	91.0	68	2025	18.2	82	3
## 377	31.0	4	91.0	68	1970	17.6	82	3
## 378	38.0	4	105.0	63	2125	14.7	82	1
## 379	36.0	4	98.0	70	2125	17.3	82	1
## 380	36.0	4	120.0	88	2160	14.5	82	3
## 381	36.0	4	107.0	75	2205	14.5	82	3
## 382	34.0	4	108.0	70	2245	16.9	82	3
## 383	38.0	4	91.0	67	1965	15.0	82	3
## 384	32.0	4	91.0	67	1965	15.7	82	3
## 385	38.0	4	91.0	67	1995	16.2	82	3
## 386	25.0	6	181.0	110	2945	16.4	82	1
## 387	38.0	6	262.0	85	3015	17.0	82	1
## 388	26.0	4	156.0	92	2585	14.5	82	1
## 389	22.0	6	232.0	112	2835	14.7	82	1
## 390	32.0	4	144.0	96	2665	13.9	82	3
## 391	36.0	4	135.0	84	2370	13.0	82	1
## 392	27.0	4	151.0	90	2950	17.3	82	1
## 393	27.0	4	140.0	86	2790	15.6	82	1
## 394	44.0	4	97.0	52	2130	24.6	82	2
## 395	32.0	4	135.0	84	2295	11.6	82	1
## 396	28.0	4	120.0	79	2625	18.6	82	1
## 397	31.0	4	119.0	82	2720	19.4	82	1
##					name mpg01			
## 1			chevrolet chevelle malibu	0				
## 2			buick skylark 320	0				
## 3			plymouth satellite	0				
## 4			amc rebel sst	0				
## 5			ford torino	0				
## 6			ford galaxie 500	0				
## 7			chevrolet impala	0				
## 8			plymouth fury iii	0				
## 9			pontiac catalina	0				
## 10			amc ambassador dpl	0				
## 11			dodge challenger se	0				
## 12			plymouth 'cuda 340	0				
## 13			chevrolet monte carlo	0				
## 14			buick estate wagon (sw)	0				
## 15			toyota corona mark ii	1				
## 16			plymouth duster	0				
## 17			amc hornet	0				
## 18			ford maverick	0				
## 19			datsum pl510	1				
## 20			volkswagen 1131 deluxe sedan	1				
## 21			peugeot 504	1				
## 22			audi 100 ls	1				
## 23			saab 99e	1				
## 24			bmw 2002	1				
## 25			amc gremlin	0				
## 26			ford f250	0				
## 27			chevy c20	0				
## 28			dodge d200	0				
## 29			hi 1200d	0				
## 30			datsum pl510	1				

## 31	chevrolet vega 2300	1
## 32	toyota corona	1
## 34	amc gremlin	0
## 35	plymouth satellite custom	0
## 36	chevrolet chevelle malibu	0
## 37	ford torino 500	0
## 38	amc matador	0
## 39	chevrolet impala	0
## 40	pontiac catalina brougham	0
## 41	ford galaxie 500	0
## 42	plymouth fury iii	0
## 43	dodge monaco (sw)	0
## 44	ford country squire (sw)	0
## 45	pontiac safari (sw)	0
## 46	amc hornet sportabout (sw)	0
## 47	chevrolet vega (sw)	0
## 48	pontiac firebird	0
## 49	ford mustang	0
## 50	mercury capri 2000	1
## 51	opel 1900	1
## 52	peugeot 304	1
## 53	fiat 124b	1
## 54	toyota corolla 1200	1
## 55	datsum 1200	1
## 56	volkswagen model 111	1
## 57	plymouth cricket	1
## 58	toyota corona hardtop	1
## 59	dodge colt hardtop	1
## 60	volkswagen type 3	1
## 61	chevrolet vega	0
## 62	ford pinto runabout	0
## 63	chevrolet impala	0
## 64	pontiac catalina	0
## 65	plymouth fury iii	0
## 66	ford galaxie 500	0
## 67	amc ambassador sst	0
## 68	mercury marquis	0
## 69	buick lesabre custom	0
## 70	oldsmobile delta 88 royale	0
## 71	chrysler newport royal	0
## 72	mazda rx2 coupe	0
## 73	amc matador (sw)	0
## 74	chevrolet chevelle concours (sw)	0
## 75	ford gran torino (sw)	0
## 76	plymouth satellite custom (sw)	0
## 77	volvo 145e (sw)	0
## 78	volkswagen 411 (sw)	0
## 79	peugeot 504 (sw)	0
## 80	renault 12 (sw)	1
## 81	ford pinto (sw)	0
## 82	datsum 510 (sw)	1
## 83	toyouta corona mark ii (sw)	1
## 84	dodge colt (sw)	1
## 85	toyota corolla 1600 (sw)	1

## 86	buick century 350	0
## 87	amc matador	0
## 88	chevrolet malibu	0
## 89	ford gran torino	0
## 90	dodge coronet custom	0
## 91	mercury marquis brougham	0
## 92	chevrolet caprice classic	0
## 93	ford ltd	0
## 94	plymouth fury gran sedan	0
## 95	chrysler new yorker brougham	0
## 96	buick electra 225 custom	0
## 97	amc ambassador brougham	0
## 98	plymouth valiant	0
## 99	chevrolet nova custom	0
## 100	amc hornet	0
## 101	ford maverick	0
## 102	plymouth duster	1
## 103	volkswagen super beetle	1
## 104	chevrolet impala	0
## 105	ford country	0
## 106	plymouth custom suburb	0
## 107	oldsmobile vista cruiser	0
## 108	amc gremlin	0
## 109	toyota carina	0
## 110	chevrolet vega	0
## 111	datsum 610	0
## 112	maxda rx3	0
## 113	ford pinto	0
## 114	mercury capri v6	0
## 115	fiat 124 sport coupe	1
## 116	chevrolet monte carlo s	0
## 117	pontiac grand prix	0
## 118	fiat 128	1
## 119	opel manta	1
## 120	audi 100ls	0
## 121	volvo 144ea	0
## 122	dodge dart custom	0
## 123	saab 99le	1
## 124	toyota mark ii	0
## 125	oldsmobile omega	0
## 126	plymouth duster	0
## 128	amc hornet	0
## 129	chevrolet nova	0
## 130	datsum b210	1
## 131	ford pinto	1
## 132	toyota corolla 1200	1
## 133	chevrolet vega	1
## 134	chevrolet chevelle malibu classic	0
## 135	amc matador	0
## 136	plymouth satellite sebring	0
## 137	ford gran torino	0
## 138	buick century luxus (sw)	0
## 139	dodge coronet custom (sw)	0
## 140	ford gran torino (sw)	0

## 141	amc matador (sw)	0
## 142	audi fox	1
## 143	volkswagen dasher	1
## 144	opel manta	1
## 145	toyota corona	1
## 146	datsum 710	1
## 147	dodge colt	1
## 148	fiat 128	1
## 149	fiat 124 tc	1
## 150	honda civic	1
## 151	subaru	1
## 152	fiat x1.9	1
## 153	plymouth valiant custom	0
## 154	chevrolet nova	0
## 155	mercury monarch	0
## 156	ford maverick	0
## 157	pontiac catalina	0
## 158	chevrolet bel air	0
## 159	plymouth grand fury	0
## 160	ford ltd	0
## 161	buick century	0
## 162	chevroelt chevelle malibu	0
## 163	amc matador	0
## 164	plymouth fury	0
## 165	buick skyhawk	0
## 166	chevrolet monza 2+2	0
## 167	ford mustang ii	0
## 168	toyota corolla	1
## 169	ford pinto	1
## 170	amc gremlin	0
## 171	pontiac astro	1
## 172	toyota corona	1
## 173	volkswagen dasher	1
## 174	datsum 710	1
## 175	ford pinto	0
## 176	volkswagen rabbit	1
## 177	amc pacer	0
## 178	audi 100ls	1
## 179	peugeot 504	1
## 180	volvo 244dl	0
## 181	saab 99le	1
## 182	honda civic cvcc	1
## 183	fiat 131	1
## 184	opel 1900	1
## 185	capri ii	1
## 186	dodge colt	1
## 187	renault 12tl	1
## 188	chevrolet chevelle malibu classic	0
## 189	dodge coronet brougham	0
## 190	amc matador	0
## 191	ford gran torino	0
## 192	plymouth valiant	0
## 193	chevrolet nova	0
## 194	ford maverick	1

## 195	amc hornet	0
## 196	chevrolet chevette	1
## 197	chevrolet woody	1
## 198	vw rabbit	1
## 199	honda civic	1
## 200	dodge aspen se	0
## 201	ford granada ghia	0
## 202	pontiac ventura sj	0
## 203	amc pacer d/l	0
## 204	volkswagen rabbit	1
## 205	datsum b-210	1
## 206	toyota corolla	1
## 207	ford pinto	1
## 208	volvo 245	0
## 209	plymouth volare premier v8	0
## 210	peugeot 504	0
## 211	toyota mark ii	0
## 212	mercedes-benz 280s	0
## 213	cadillac seville	0
## 214	chevy c10	0
## 215	ford f108	0
## 216	dodge d100	0
## 217	honda accord cvcc	1
## 218	buick opel isuzu deluxe	1
## 219	renault 5 gtl	1
## 220	plymouth arrow gs	1
## 221	datsum f-10 hatchback	1
## 222	chevrolet caprice classic	0
## 223	oldsmobile cutlass supreme	0
## 224	dodge monaco brougham	0
## 225	mercury cougar brougham	0
## 226	chevrolet concours	0
## 227	buick skylark	0
## 228	plymouth volare custom	0
## 229	ford granada	0
## 230	pontiac grand prix lj	0
## 231	chevrolet monte carlo landau	0
## 232	chrysler cordoba	0
## 233	ford thunderbird	0
## 234	volkswagen rabbit custom	1
## 235	pontiac sunbird coupe	1
## 236	toyota corolla liftback	1
## 237	ford mustang ii 2+2	1
## 238	chevrolet chevette	1
## 239	dodge colt m/m	1
## 240	subaru dl	1
## 241	volkswagen dasher	1
## 242	datsum 810	0
## 243	bmw 320i	0
## 244	mazda rx-4	0
## 245	volkswagen rabbit custom diesel	1
## 246	ford fiesta	1
## 247	mazda glc deluxe	1
## 248	datsum b210 gx	1

## 249	honda civic cvcc	1
## 250	oldsmobile cutlass salon brougham	0
## 251	dodge diplomat	0
## 252	mercury monarch ghia	0
## 253	pontiac phoenix lj	0
## 254	chevrolet malibu	0
## 255	ford fairmont (auto)	0
## 256	ford fairmont (man)	1
## 257	plymouth volare	0
## 258	amc concord	0
## 259	buick century special	0
## 260	mercury zephyr	0
## 261	dodge aspen	0
## 262	amc concord d/l	0
## 263	chevrolet monte carlo landau	0
## 264	buick regal sport coupe (turbo)	0
## 265	ford futura	0
## 266	dodge magnum xe	0
## 267	chevrolet chevette	1
## 268	toyota corona	1
## 269	datsum 510	1
## 270	dodge omni	1
## 271	toyota celica gt liftback	0
## 272	plymouth sapporo	1
## 273	oldsmobile starfire sx	1
## 274	datsum 200-sx	1
## 275	audi 5000	0
## 276	volvo 264gl	0
## 277	saab 99gle	0
## 278	peugeot 604sl	0
## 279	volkswagen scirocco	1
## 280	honda accord lx	1
## 281	pontiac lemans v6	0
## 282	mercury zephyr 6	0
## 283	ford fairmont 4	0
## 284	amc concord dl 6	0
## 285	dodge aspen 6	0
## 286	chevrolet caprice classic	0
## 287	ford ltd landau	0
## 288	mercury grand marquis	0
## 289	dodge st. regis	0
## 290	buick estate wagon (sw)	0
## 291	ford country squire (sw)	0
## 292	chevrolet malibu classic (sw)	0
## 293	chrysler lebaron town @ country (sw)	0
## 294	vw rabbit custom	1
## 295	maxda glc deluxe	1
## 296	dodge colt hatchback custom	1
## 297	amc spirit dl	1
## 298	mercedes benz 300d	1
## 299	cadillac eldorado	1
## 300	peugeot 504	1
## 301	oldsmobile cutlass salon brougham	1
## 302	plymouth horizon	1

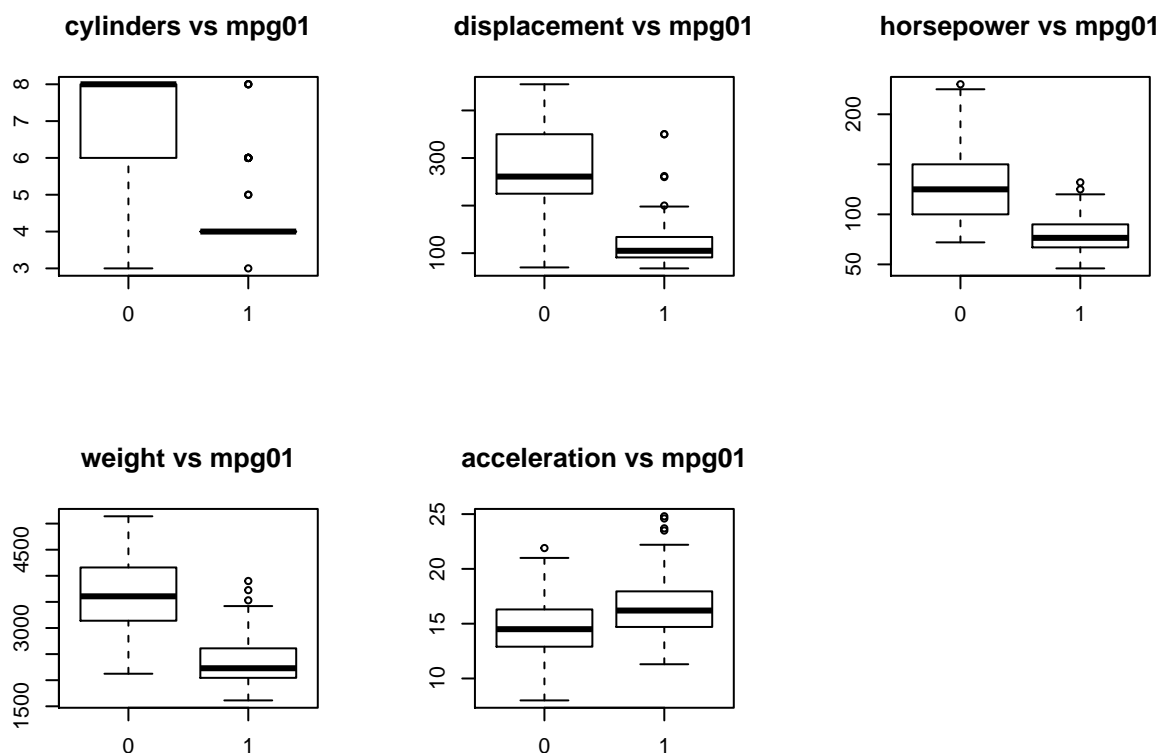
## 303	plymouth horizon tc3	1
## 304	datsun 210	1
## 305	fiat strada custom	1
## 306	buick skylark limited	1
## 307	chevrolet citation	1
## 308	oldsmobile omega brougham	1
## 309	pontiac phoenix	1
## 310	vw rabbit	1
## 311	toyota corolla tercel	1
## 312	chevrolet chevette	1
## 313	datsun 310	1
## 314	chevrolet citation	1
## 315	ford fairmont	1
## 316	amc concord	1
## 317	dodge aspen	0
## 318	audi 4000	1
## 319	toyota corona liftback	1
## 320	mazda 626	1
## 321	datsun 510 hatchback	1
## 322	toyota corolla	1
## 323	mazda glc	1
## 324	dodge colt	1
## 325	datsun 210	1
## 326	vw rabbit c (diesel)	1
## 327	vw dasher (diesel)	1
## 328	audi 5000s (diesel)	1
## 329	mercedes-benz 240d	1
## 330	honda civic 1500 gl	1
## 332	subaru dl	1
## 333	volkswagen rabbit	1
## 334	datsun 280-zx	1
## 335	mazda rx-7 gs	1
## 336	triumph tr7 coupe	1
## 338	honda accord	1
## 339	plymouth reliant	1
## 340	buick skylark	1
## 341	dodge aries wagon (sw)	1
## 342	chevrolet citation	1
## 343	plymouth reliant	1
## 344	toyota starlet	1
## 345	plymouth champ	1
## 346	honda civic 1300	1
## 347	subaru	1
## 348	datsun 210 mpg	1
## 349	toyota tercel	1
## 350	mazda glc 4	1
## 351	plymouth horizon 4	1
## 352	ford escort 4w	1
## 353	ford escort 2h	1
## 354	volkswagen jetta	1
## 356	honda prelude	1
## 357	toyota corolla	1
## 358	datsun 200sx	1
## 359	mazda 626	1

## 360	peugeot 505s turbo diesel	1
## 361	volvo diesel	1
## 362	toyota cressida	1
## 363	datsum 810 maxima	1
## 364	buick century	0
## 365	oldsmobile cutlass ls	1
## 366	ford granada gl	0
## 367	chrysler lebaron salon	0
## 368	chevrolet cavalier	1
## 369	chevrolet cavalier wagon	1
## 370	chevrolet cavalier 2-door	1
## 371	pontiac j2000 se hatchback	1
## 372	dodge aries se	1
## 373	pontiac phoenix	1
## 374	ford fairmont futura	1
## 375	volkswagen rabbit l	1
## 376	mazda glc custom l	1
## 377	mazda glc custom	1
## 378	plymouth horizon miser	1
## 379	mercury lynx l	1
## 380	nissan stanza xe	1
## 381	honda accord	1
## 382	toyota corolla	1
## 383	honda civic	1
## 384	honda civic (auto)	1
## 385	datsum 310 gx	1
## 386	buick century limited	1
## 387	oldsmobile cutlass ciera (diesel)	1
## 388	chrysler lebaron medallion	1
## 389	ford granada l	0
## 390	toyota celica gt	1
## 391	dodge charger 2.2	1
## 392	chevrolet camaro	1
## 393	ford mustang gl	1
## 394	vw pickup	1
## 395	dodge rampage	1
## 396	ford ranger	1
## 397	chevy s-10	1

(b)

```
par(mfrow=c(2,3))
name=c("cylinders","displacement","horsepower","weight","acceleration")

boxplot(cylinders ~ mpg01, data = mpg01, main = "cylinders vs mpg01")
boxplot(displacement ~ mpg01, data = mpg01, main = "displacement vs mpg01")
boxplot(horsepower ~ mpg01, data = mpg01, main = "horsepower vs mpg01")
boxplot(weight ~ mpg01, data = mpg01, main = "weight vs mpg01")
boxplot(acceleration ~ mpg01, data = mpg01, main = "acceleration vs mpg01")
```



#(c)

```
set.seed(1)
rows <- sample(x=nrow(mpg01), size=0.8*nrow(mpg01))
rows
```

```
## [1] 105 146 224 354 79 348 365 255 242 24 388 68 262 391 292 188 270
## [18] 372 143 290 387 382 384 47 99 142 5 140 317 124 175 217 178 67
## [35] 297 239 283 39 257 379 289 228 275 194 185 274 9 165 252 238 164
## [52] 294 149 83 383 34 107 174 222 136 304 98 152 110 214 85 157 250
## [69] 28 356 329 376 111 336 330 323 347 123 245 301 333 334 363 101 234
## [86] 63 218 38 75 44 73 18 193 263 233 237 135 121 357 360 192 103
## [103] 371 287 183 62 305 137 299 170 276 206 100 295 42 4 198 29 315
## [120] 362 321 296 131 369 203 122 312 55 61 326 151 21 10 167 240 154
## [137] 144 271 251 129 173 380 60 65 181 112 303 288 26 211 340 385 373
## [154] 109 120 43 125 313 249 50 359 207 291 179 201 94 15 76 163 225
## [171] 386 186 189 86 339 195 311 160 130 300 307 41 187 106 314 40 284
## [188] 370 213 247 256 258 261 375 57 117 22 342 352 318 52 278 368 51
## [205] 35 97 392 338 48 132 273 19 138 364 353 265 115 259 166 59 46
## [222] 350 177 87 156 219 358 8 69 216 282 196 325 309 349 341 108 169
## [239] 232 70 269 88 285 161 158 32 212 20 389 241 281 208 66 84 202
## [256] 226 337 381 298 236 153 191 37 102 90 200 346 95 77 127 345 14
## [273] 172 316 36 23 30 119 229 254 277 344 210 243 6 150 377 17 279
## [290] 197 199 104 190 33 134 180 260 12 366 155 148 126 45 272 266 93
## [307] 16 141 168 113 133 182 374
```

```
trainset <- mpg01[rows, ]
testset <- mpg01[-rows, ]
```


(d)

```
lda.fit <- lda(mpg01 ~ displacement+horsepower+weight+acceleration+year+cylinders+origin, data=trainset)
lda.pred <- predict(lda.fit, testset)
table(testset$mpg01, lda.pred$class)

##
##      0  1
##    0 30  5
##    1  0 44
# accuracy of 0=100%, accuracy of 1=44/(44+5)
```

(e)

```
qda.fit <- qda(mpg01 ~ displacement+horsepower+weight+acceleration+year+cylinders+origin, data=trainset)
qda.pred <- predict(qda.fit, testset)
table(testset$mpg01, qda.pred$class)

##
##      0  1
##    0 32  3
##    1  2 42
# accuracy of 0=32/(32+2), accuracy of 1=42/(3+42)
```

(f)

```
modf <- glm(as.factor(mpg01) ~ displacement+horsepower+weight+acceleration+year+cylinders+origin, data=
lr.probs <- predict(modf, testset, type="response")
lr.pred <- ifelse(lr.probs>0.5, "1", "0")
table(testset$mpg01, lr.pred)

##      lr.pred
##      0  1
##    0 30  5
##    1  1 43
# accuracy of 0=30/(30+1), accuracy of 1=43/(5+43)
```

(g)

4.12 (a)

```
Power <- function(){ print( 2^3)}
Power()

## [1] 8
```

(b)

```
Power2 <- function(x,a){  
  print( x^a)  
}
```

```
Power2(3,8)
```

```
## [1] 6561
```

(c)

```
Power2(10,3)
```

```
## [1] 1000
```

```
Power2(8,17)
```

```
## [1] 2.2518e+15
```

```
Power2(131,3)
```

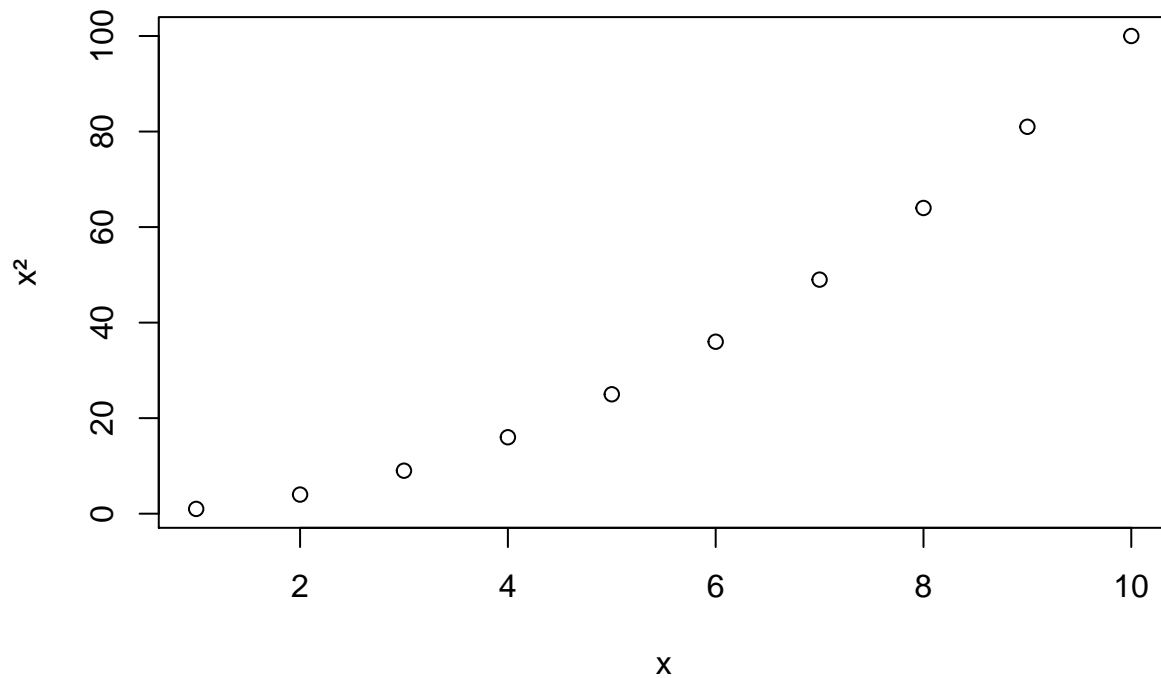
```
## [1] 2248091
```

(d)

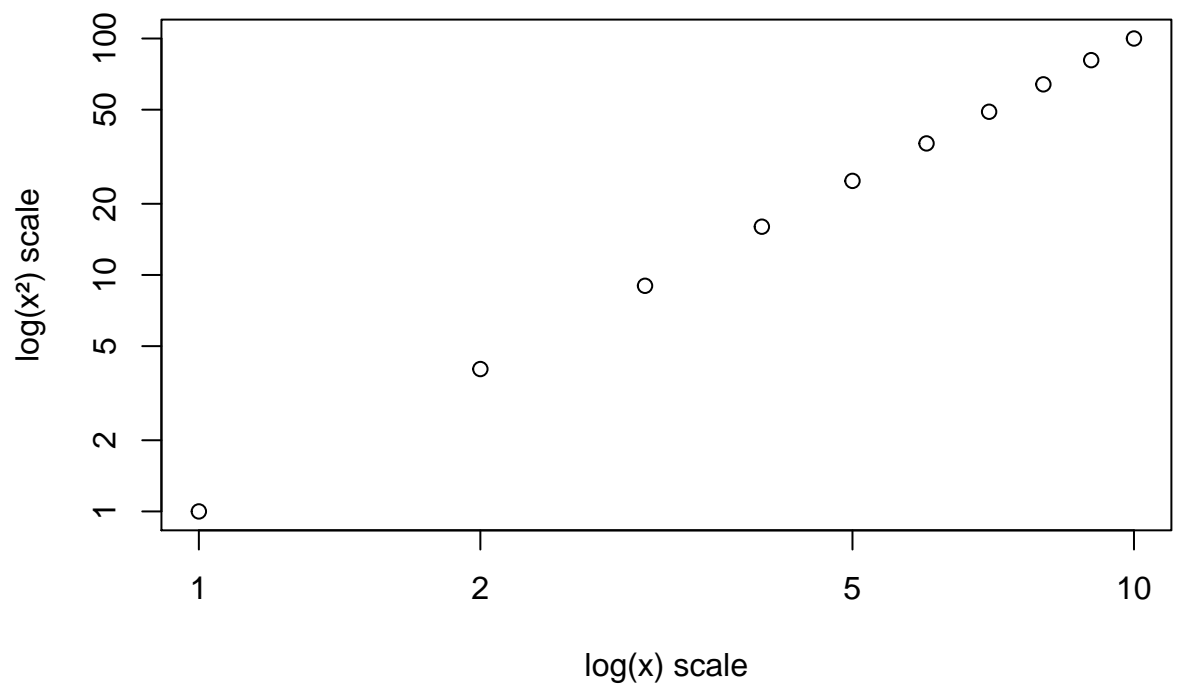
```
Power3 <- function(x,a){  
  return( x^a)  
}
```

(f)

```
plot(x = x<-1:10, y= y<-Power3(x,2), xlab="x", ylab="x2")
```



```
plot(x,y,log="xy", xlab="log(x) scale", ylab="log(x2) scale")
```



#(g)

```
PlotPower <- function(x,a){
  plot(x = x, y= y<-Power3(x,a), xlab="x", ylab=paste0("x^",a))
}
```

```
PlotPower(1:10,3)
```

