Lab 4.7

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```
library(ISLR2)
```

4.7.1

```
## Warning: package 'ISLR2' was built under R version 4.3.2
```

```
names(Smarket)
```

```
## [1] "Year"      "Lag1"      "Lag2"      "Lag3"      "Lag4"      "Lag5"
## [7] "Volume"    "Today"     "Direction"
```

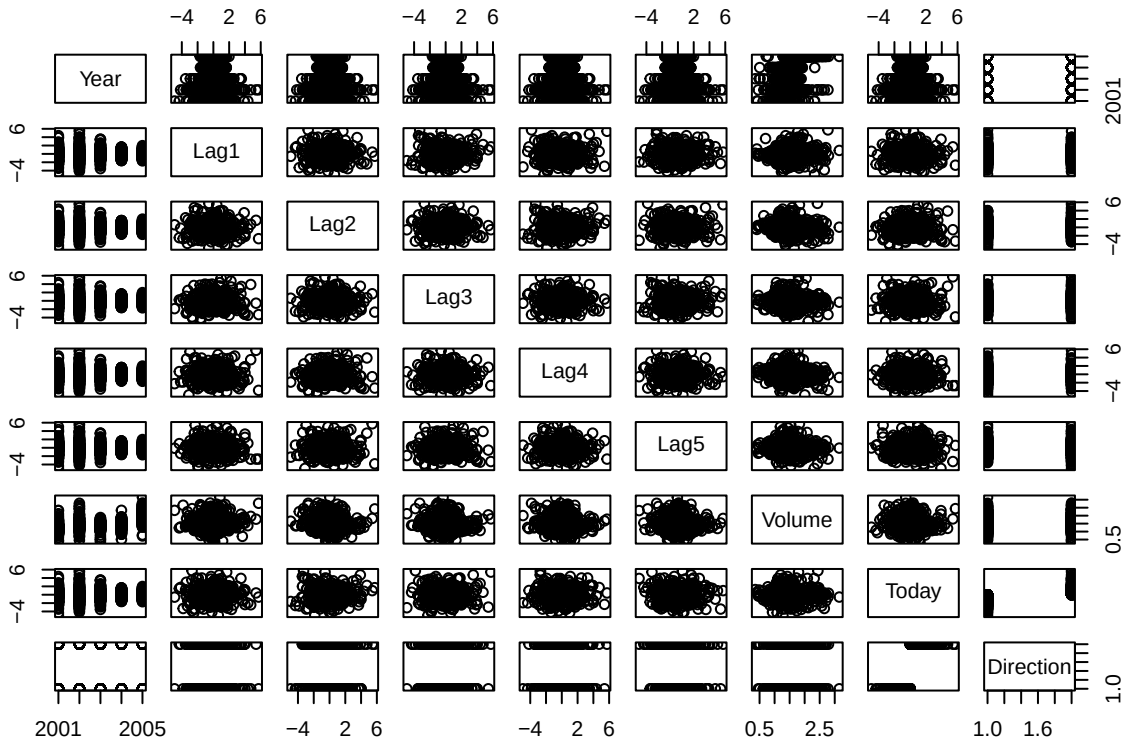
```
dim(Smarket)
```

```
## [1] 1250    9
```

```
summary(Smarket)
```

```
##       Year           Lag1                Lag2                Lag3          
##  Min.   :2001   Min.   :-4.922000   Min.   :-4.922000   Min.   :-4.922000  
##  1st Qu.:2002   1st Qu.:-0.639500   1st Qu.:-0.639500   1st Qu.:-0.640000  
##  Median :2003   Median : 0.039000   Median : 0.039000   Median : 0.038500  
##  Mean   :2003   Mean   : 0.003834   Mean   : 0.003919   Mean   : 0.001716  
##  3rd Qu.:2004   3rd Qu.: 0.596750   3rd Qu.: 0.596750   3rd Qu.: 0.596750  
##  Max.   :2005   Max.   : 5.733000   Max.   : 5.733000   Max.   : 5.733000  
##       Lag4                Lag5              Volume           Today          
##  Min.   :-4.922000   Min.   :-4.92200   Min.   :0.3561   Min.   :-4.922000  
##  1st Qu.:-0.640000   1st Qu.:-0.64000   1st Qu.:1.2574   1st Qu.:-0.639500  
##  Median : 0.038500   Median : 0.03850   Median :1.4229   Median : 0.038500  
##  Mean   : 0.001636   Mean   : 0.00561   Mean   :1.4783   Mean   : 0.003138  
##  3rd Qu.: 0.596750   3rd Qu.: 0.59700   3rd Qu.:1.6417   3rd Qu.: 0.596750  
##  Max.   : 5.733000   Max.   : 5.73300   Max.   :3.1525   Max.   : 5.733000  
##  Direction
##  Down:602
##  Up  :648
##
##
##
##
```

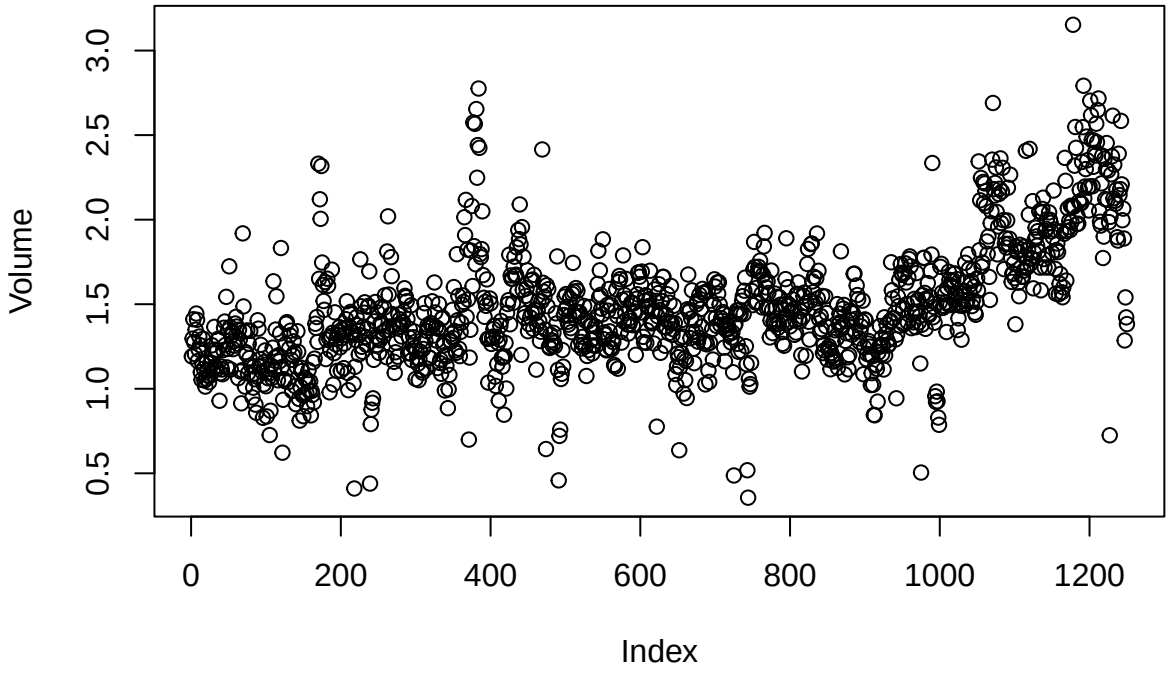
```
pairs(Smarket)
```




```
#cor(Smarket)
cor(Smarket[, -9])
```

```
##              Year         Lag1         Lag2         Lag3         Lag4
## Year    1.00000000  0.029699649  0.030596422  0.033194581  0.035688718
## Lag1    0.02969965  1.000000000 -0.026294328 -0.010803402 -0.002985911
## Lag2    0.03059642 -0.026294328  1.000000000 -0.025896670 -0.010853533
## Lag3    0.03319458 -0.010803402 -0.025896670  1.000000000 -0.024051036
## Lag4    0.03568872 -0.002985911 -0.010853533 -0.024051036  1.000000000
## Lag5    0.02978799 -0.005674606 -0.003557949 -0.018808338 -0.027083641
## Volume 0.53900647  0.040909908 -0.043383215 -0.041823686 -0.048414246
## Today  0.03009523 -0.026155045 -0.010250033 -0.002447647 -0.006899527
##                Lag5       Volume        Today
## Year     0.029787995  0.53900647  0.030095229
## Lag1    -0.005674606  0.04090991 -0.026155045
## Lag2    -0.003557949 -0.04338321 -0.010250033
## Lag3    -0.018808338 -0.04182369 -0.002447647
## Lag4    -0.027083641 -0.04841425 -0.006899527
## Lag5     1.000000000 -0.02200231 -0.034860083
## Volume -0.022002315  1.00000000  0.014591823
## Today  -0.034860083  0.01459182  1.000000000
```

```r
attach(Smarket)
plot(Volume)
```




```r
glm.fits <- glm(
Direction ~ Lag1 + Lag2 + Lag3 + Lag4 + Lag5 + Volume ,
data = Smarket , family = binomial
)
summary(glm.fits)
```

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```
##
## Call:
## glm(formula = Direction ~ Lag1 + Lag2 + Lag3 + Lag4 + Lag5 +
##     Volume, family = binomial, data = Smarket)
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept) -0.126000   0.240736  -0.523    0.601
## Lag1        -0.073074   0.050167  -1.457    0.145
## Lag2        -0.042301   0.050086  -0.845    0.398
## Lag3         0.011085   0.049939   0.222    0.824
```

```
## Lag4         0.009359   0.049974   0.187    0.851
## Lag5         0.010313   0.049511   0.208    0.835
## Volume       0.135441   0.158360   0.855    0.392
##
## (Dispersion parameter for binomial family taken to be 1)
##
##     Null deviance: 1731.2  on 1249  degrees of freedom
## Residual deviance: 1727.6  on 1243  degrees of freedom
## AIC: 1741.6
##
## Number of Fisher Scoring iterations: 3
```

```
coef(glm.fits)
```

```
##  (Intercept)         Lag1         Lag2         Lag3         Lag4         Lag5
## -0.126000257 -0.073073746 -0.042301344  0.011085108  0.009358938  0.010313068
##       Volume
##  0.135440659
```

```
summary(glm.fits)$coef
```

```
##                 Estimate Std. Error    z value  Pr(>|z|)
## (Intercept) -0.126000257 0.24073574 -0.5233966 0.6006983
## Lag1        -0.073073746 0.05016739 -1.4565986 0.1452272
## Lag2        -0.042301344 0.05008605 -0.8445733 0.3983491
## Lag3         0.011085108 0.04993854  0.2219750 0.8243333
## Lag4         0.009358938 0.04997413  0.1872757 0.8514445
## Lag5         0.010313068 0.04951146  0.2082966 0.8349974
## Volume       0.135440659 0.15835970  0.8552723 0.3924004
```

```
summary(glm.fits)$coef[, 4]
```

```
## (Intercept)        Lag1        Lag2        Lag3        Lag4        Lag5
##   0.6006983   0.1452272   0.3983491   0.8243333   0.8514445   0.8349974
##      Volume
##   0.3924004
```

```
glm.probs <- predict(glm.fits , type = "response")
glm.probs[1:10]
```

```
##         1         2         3         4         5         6         7         8
## 0.5070841 0.4814679 0.4811388 0.5152224 0.5107812 0.5069565 0.4926509 0.5092292
##         9        10
## 0.5176135 0.4888378
```

```
contrasts(Direction)
```

```
##      Up
## Down  0
## Up    1
```

```
glm.pred <- rep("Down", 1250)
glm.pred[glm.probs > .5] = "Up"

table(glm.pred , Direction)
```

```
##         Direction
## glm.pred Down  Up
##     Down  145 141
##     Up    457 507
```

```
(507 + 145) / 1250
```

```
## [1] 0.5216
```

```
mean(glm.pred == Direction)
```

```
## [1] 0.5216
```

```
train <- (Year < 2005)
Smarket.2005 <- Smarket[!train, ]
dim(Smarket.2005)
```

```
## [1] 252   9
```

```
Direction.2005 <- Direction[!train]

glm.fits <- glm(
Direction ~ Lag1 + Lag2 + Lag3 + Lag4 + Lag5 + Volume,
data = Smarket, family = binomial, subset = train
)

glm.probs <- predict(glm.fits, Smarket.2005,
type = "response")

glm.pred <- rep("Down", 252)
glm.pred[glm.probs > .5] <- "Up"
table(glm.pred , Direction.2005)
```

```
##         Direction.2005
## glm.pred Down Up
##     Down   77 97
##     Up     34 44
```

```
mean(glm.pred == Direction.2005)
```

```
## [1] 0.4801587
```

```
mean(glm.pred != Direction.2005)
```

```
## [1] 0.5198413
```

```
glm.fits <- glm(Direction ~ Lag1 + Lag2,
  data = Smarket,family = binomial, subset = train)
glm.probs <- predict(glm.fits, Smarket.2005,
  type = "response")
glm.pred <- rep("Down", 252)
glm.pred[glm.probs > .5] <- "Up"
table(glm.pred , Direction.2005)
```

```
##         Direction.2005
## glm.pred Down  Up
##     Down   35  35
##     Up     76 106
```

```
mean(glm.pred == Direction.2005)
```

```
## [1] 0.5595238
```

```
106 / (106 + 76)
```

```
## [1] 0.5824176
```

```
predict(glm.fits,
  newdata =
    data.frame(Lag1 = c(1.2, 1.5), Lag2 = c(1.1, -0.8)),
  type = "response"
)
```

```
##         1         2
## 0.4791462 0.4960939
```

```
library(MASS)
```

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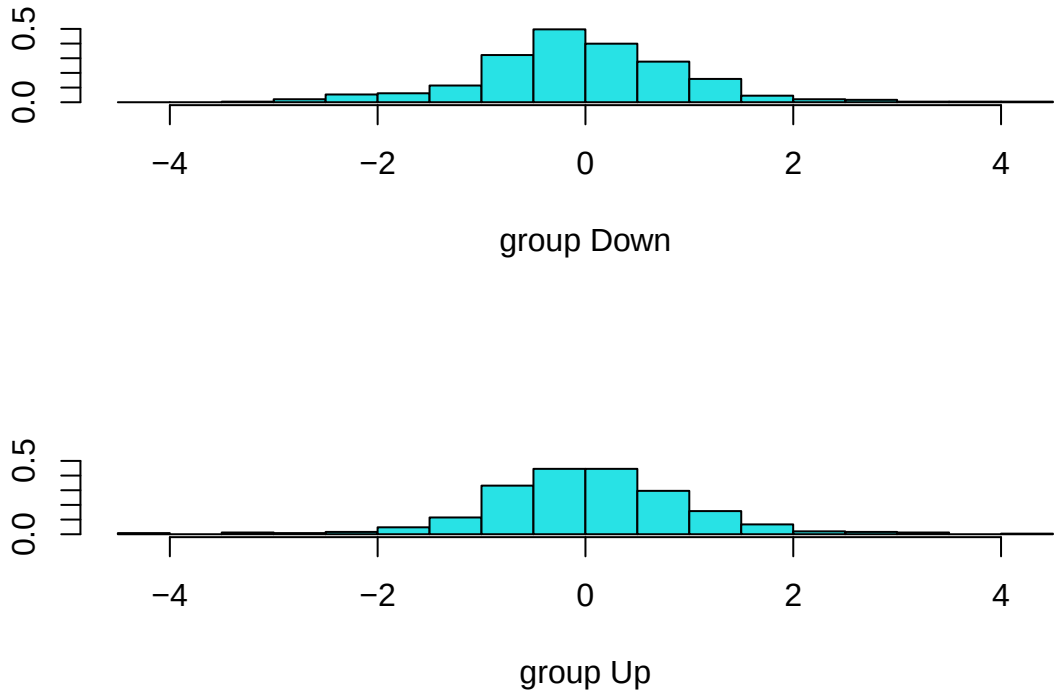
```
##
## Attaching package: 'MASS'
```

```
## The following object is masked from 'package:ISLR2':
##
##     Boston
```

```
lda.fit <- lda(Direction ~ Lag1 + Lag2, data = Smarket, subset = train)
lda.fit
```

```
## Call:
## lda(Direction ~ Lag1 + Lag2, data = Smarket, subset = train)
##
## Prior probabilities of groups:
##     Down       Up
## 0.491984 0.508016
##
## Group means:
##             Lag1        Lag2
## Down  0.04279022  0.03389409
## Up   -0.03954635 -0.03132544
##
## Coefficients of linear discriminants:
##             LD1
## Lag1 -0.6420190
## Lag2 -0.5135293
```

```
plot(lda.fit)
```




```
lda.pred <- predict(lda.fit, Smarket.2005)
names(lda.pred)
```

```
## [1] "class"     "posterior" "x"
```

```
lda.class <- lda.pred$class
table(lda.class, Direction.2005)
```

```
##          Direction.2005
## lda.class Down  Up
##      Down   35  35
##      Up     76 106
```

```
sum(lda.pred$posterior[, 1] >= .5)
```

```
## [1] 70
```

```
sum(lda.pred$posterior[, 1] < .5)
```

```
## [1] 182
```

```
lda.pred$posterior[1:20, 1]
```

```
##       999      1000      1001      1002      1003      1004      1005      1006
## 0.4901792 0.4792185 0.4668185 0.4740011 0.4927877 0.4938562 0.4951016 0.4872861
##      1007      1008      1009      1010      1011      1012      1013      1014
## 0.4907013 0.4844026 0.4906963 0.5119988 0.4895152 0.4706761 0.4744593 0.4799583
##      1015      1016      1017      1018
## 0.4935775 0.5030894 0.4978806 0.4886331
```

```
lda.class [1:20]
```

```
##  [1] Up   Up   Up   Up   Up   Up   Up   Up   Up   Up   Up   Down Up   Up   Up
## [16] Up   Up   Down Up   Up
## Levels: Down Up
```

```
sum(lda.pred$posterior[, 1] > .9)
```

```
## [1] 0
```

```
qda.fit <- qda(Direction ~ Lag1 + Lag2 , data = Smarket, subset = train)
qda.fit
```

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```
## Call:
## qda(Direction ~ Lag1 + Lag2, data = Smarket, subset = train)
##
## Prior probabilities of groups:
```

```
##     Down       Up
## 0.491984 0.508016
##
## Group means:
##             Lag1        Lag2
## Down  0.04279022  0.03389409
## Up   -0.03954635 -0.03132544
```

```
qda.class <- predict(qda.fit, Smarket.2005)$class
table(qda.class , Direction.2005)
```

```
##          Direction.2005
## qda.class Down  Up
##      Down   30  20
##      Up     81 121
```

```
mean(qda.class == Direction.2005)
```

```
## [1] 0.5992063
```

```
library(e1071)
```

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```
## Warning: package 'e1071' was built under R version 4.3.2
```

```
nb.fit <- naiveBayes(Direction ~ Lag1 + Lag2, data = Smarket, subset = train)
nb.fit
```

```
##
## Naive Bayes Classifier for Discrete Predictors
##
## Call:
## naiveBayes.default(x = X, y = Y, laplace = laplace)
##
## A-priori probabilities:
## Y
##     Down       Up
## 0.491984 0.508016
##
## Conditional probabilities:
##       Lag1
## Y             [,1]     [,2]
##   Down  0.04279022 1.227446
##   Up   -0.03954635 1.231668
##
##       Lag2
## Y             [,1]     [,2]
##   Down  0.03389409 1.239191
##   Up   -0.03132544 1.220765
```

```
mean(Lag1[train][Direction[train] == "Down"])
```

```
## [1] 0.04279022
```

```
sd(Lag1[train][Direction[train] == "Down"])
```

```
## [1] 1.227446
```

```
nb.class <- predict(nb.fit, Smarket.2005)
table(nb.class, Direction.2005)
```

```
##         Direction.2005
## nb.class Down  Up
##     Down   28  20
##     Up     83 121
```

```
mean(nb.class == Direction.2005)
```

```
## [1] 0.5912698
```

```
nb.preds <- predict(nb.fit, Smarket.2005, type = "raw")
nb.preds[1:5, ]
```

```
##           Down        Up
## [1,] 0.4873164 0.5126836
## [2,] 0.4762492 0.5237508
## [3,] 0.4653377 0.5346623
## [4,] 0.4748652 0.5251348
## [5,] 0.4901890 0.5098110
```

```
library(class)
train.X <- cbind(Lag1 , Lag2)[train , ]
test.X <- cbind(Lag1 , Lag2)[!train , ]
train.Direction <- Direction[train]

set.seed(1)
knn.pred <- knn(train.X, test.X, train.Direction, k = 1)
table(knn.pred , Direction.2005)
```

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```
##         Direction.2005
## knn.pred Down Up
##     Down   43 58
##     Up     68 83
```

```
(83 + 43) / 252
```

```
## [1] 0.5
```

```
knn.pred <- knn(train.X, test.X, train.Direction, k = 3)
table(knn.pred , Direction.2005)
```

```
##         Direction.2005
## knn.pred Down Up
##     Down   48 54
##     Up     63 87
```

```
mean(knn.pred == Direction.2005)
```

```
## [1] 0.5357143
```

```
dim(Caravan)
```

```
## [1] 5822   86
```

```
attach(Caravan)
summary(Purchase)
```

```
##   No  Yes
## 5474  348
```

```
348 / 5822
```

```
## [1] 0.05977327
```

```
standardized.X <- scale(Caravan[, -86])
var(Caravan[, 1])
```

```
## [1] 165.0378
```

```
var(Caravan[, 2])
```

```
## [1] 0.1647078
```

```
var(standardized.X[, 1])
```

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

```
var(standardized.X[, 2])
```

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

```
test <- 1:1000
train.X <- standardized.X[-test, ]
test.X <- standardized.X[test, ]
train.Y <- Purchase[-test]
test.Y <- Purchase[test]
set.seed(1)
knn.pred <- knn(train.X, test.X, train.Y, k = 1)
mean(test.Y != knn.pred)
```

```
## [1] 0.118
```

```
mean(test.Y != "No")
```

```
## [1] 0.059
```

```
table(knn.pred , test.Y)
```

```
##         test.Y
## knn.pred  No Yes
##      No  873  50
##      Yes  68   9
```

```
9 / (68 + 9)
```

```
## [1] 0.1168831
```

```
knn.pred <- knn(train.X, test.X, train.Y, k = 3)
table(knn.pred , test.Y)
```

```
##         test.Y
## knn.pred  No Yes
##      No  920  54
##      Yes  21   5
```

```
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```

```
## [1] 0.1923077
```

```
knn.pred <- knn(train.X, test.X, train.Y, k = 5)
table(knn.pred , test.Y)
```

```
##         test.Y
## knn.pred  No Yes
##      No  930  55
##      Yes  11   4
```

```
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```

```
## [1] 0.2666667
```

```
glm.fits <- glm(Purchase ~ ., data = Caravan, family = binomial , subset = -test)
```

```
## Warning: glm.fit: fitted probabilities numerically 0 or 1 occurred
```

```
glm.probs <- predict(glm.fits , Caravan[test , ], type = "response")
glm.pred <- rep("No", 1000)
glm.pred[glm.probs > .5] <- "Yes"
table(glm.pred , test.Y)
```

```
##         test.Y
## glm.pred  No Yes
##      No  934  59
##      Yes   7   0
```

```
glm.pred <- rep("No", 1000)
glm.pred[glm.probs > .25] <- "Yes"
table(glm.pred , test.Y)
```

```
##         test.Y
## glm.pred  No Yes
##      No  919  48
##      Yes  22  11
```

```
11 / (22 + 11)
```

```
## [1] 0.3333333
```

```
attach(Bikeshare)
dim(Bikeshare)
```

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```
## [1] 8645   15
```

```
names(Bikeshare)
```

```
##  [1] "season"     "mnth"       "day"        "hr"         "holiday"
##  [6] "weekday"    "workingday" "weathersit" "temp"       "atemp"
## [11] "hum"        "windspeed"  "casual"     "registered" "bikers"
```

```
mod.lm <- lm(
bikers ~ mnth + hr + workingday + temp + weathersit, data = Bikeshare
)
summary(mod.lm)
```

```
##
## Call:
## lm(formula = bikers ~ mnth + hr + workingday + temp + weathersit,
##     data = Bikeshare)
##
## Residuals:
##     Min      1Q  Median      3Q     Max
## -299.00  -45.70   -6.23   41.08  425.29
##
## Coefficients:
##                           Estimate Std. Error t value Pr(>|t|)
## (Intercept)               -68.632      5.307 -12.932  < 2e-16 ***
## mnthFeb                     6.845      4.287   1.597 0.110398
## mnthMarch                  16.551      4.301   3.848 0.000120 ***
## mnthApril                  41.425      4.972   8.331  < 2e-16 ***
## mnthMay                    72.557      5.641  12.862  < 2e-16 ***
## mnthJune                   67.819      6.544  10.364  < 2e-16 ***
## mnthJuly                   45.324      7.081   6.401 1.63e-10 ***
## mnthAug                    53.243      6.640   8.019 1.21e-15 ***
## mnthSept                   66.678      5.925  11.254  < 2e-16 ***
## mnthOct                    75.834      4.950  15.319  < 2e-16 ***
## mnthNov                    60.310      4.610  13.083  < 2e-16 ***
## mnthDec                    46.458      4.271  10.878  < 2e-16 ***
## hr1                       -14.579      5.699  -2.558 0.010536 *
## hr2                       -21.579      5.733  -3.764 0.000168 ***
## hr3                       -31.141      5.778  -5.389 7.26e-08 ***
## hr4                       -36.908      5.802  -6.361 2.11e-10 ***
## hr5                       -24.135      5.737  -4.207 2.61e-05 ***
## hr6                        20.600      5.704   3.612 0.000306 ***
## hr7                       120.093      5.693  21.095  < 2e-16 ***
## hr8                       223.662      5.690  39.310  < 2e-16 ***
## hr9                       120.582      5.693  21.182  < 2e-16 ***
## hr10                       83.801      5.705  14.689  < 2e-16 ***
## hr11                      105.423      5.722  18.424  < 2e-16 ***
## hr12                      137.284      5.740  23.916  < 2e-16 ***
## hr13                      136.036      5.760  23.617  < 2e-16 ***
## hr14                      126.636      5.776  21.923  < 2e-16 ***
## hr15                      132.087      5.780  22.852  < 2e-16 ***
## hr16                      178.521      5.772  30.927  < 2e-16 ***
## hr17                      296.267      5.749  51.537  < 2e-16 ***
## hr18                      269.441      5.736  46.976  < 2e-16 ***
## hr19                      186.256      5.714  32.596  < 2e-16 ***
## hr20                      125.549      5.704  22.012  < 2e-16 ***
## hr21                       87.554      5.693  15.378  < 2e-16 ***
## hr22                       59.123      5.689  10.392  < 2e-16 ***
## hr23                       26.838      5.688   4.719 2.41e-06 ***
## workingday                  1.270      1.784   0.711 0.476810
## temp                      157.209     10.261  15.321  < 2e-16 ***
```

```
## weathersitcloudy/misty     -12.890      1.964  -6.562 5.60e-11 ***
## weathersitlight rain/snow  -66.494      2.965 -22.425  < 2e-16 ***
## weathersitheavy rain/snow -109.745     76.667  -1.431 0.152341
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 76.5 on 8605 degrees of freedom
## Multiple R-squared:  0.6745, Adjusted R-squared:  0.6731
## F-statistic: 457.3 on 39 and 8605 DF,  p-value: < 2.2e-16
```

```
contrasts(Bikeshare$hr) = contr.sum(24)
contrasts(Bikeshare$mnth) = contr.sum(12)
mod.lm2 <- lm(
bikers ~ mnth + hr + workingday + temp + weathersit ,
data = Bikeshare
)
summary(mod.lm2)
```

```
##
## Call:
## lm(formula = bikers ~ mnth + hr + workingday + temp + weathersit,
##     data = Bikeshare)
##
## Residuals:
##     Min      1Q  Median      3Q     Max
## -299.00  -45.70   -6.23   41.08  425.29
##
## Coefficients:
##                            Estimate Std. Error t value Pr(>|t|)
## (Intercept)                 73.5974     5.1322  14.340  < 2e-16 ***
## mnth1                      -46.0871     4.0855 -11.281  < 2e-16 ***
## mnth2                      -39.2419     3.5391 -11.088  < 2e-16 ***
## mnth3                      -29.5357     3.1552  -9.361  < 2e-16 ***
## mnth4                       -4.6622     2.7406  -1.701  0.08895 .
## mnth5                       26.4700     2.8508   9.285  < 2e-16 ***
## mnth6                       21.7317     3.4651   6.272 3.75e-10 ***
## mnth7                       -0.7626     3.9084  -0.195  0.84530
## mnth8                        7.1560     3.5347   2.024  0.04295 *
## mnth9                       20.5912     3.0456   6.761 1.46e-11 ***
## mnth10                      29.7472     2.6995  11.019  < 2e-16 ***
## mnth11                      14.2229     2.8604   4.972 6.74e-07 ***
## hr1                        -96.1420     3.9554 -24.307  < 2e-16 ***
## hr2                       -110.7213     3.9662 -27.916  < 2e-16 ***
## hr3                       -117.7212     4.0165 -29.310  < 2e-16 ***
## hr4                       -127.2828     4.0808 -31.191  < 2e-16 ***
## hr5                       -133.0495     4.1168 -32.319  < 2e-16 ***
## hr6                       -120.2775     4.0370 -29.794  < 2e-16 ***
## hr7                        -75.5424     3.9916 -18.925  < 2e-16 ***
## hr8                         23.9511     3.9686   6.035 1.65e-09 ***
## hr9                        127.5199     3.9500  32.284  < 2e-16 ***
## hr10                        24.4399     3.9360   6.209 5.57e-10 ***
## hr11                       -12.3407     3.9361  -3.135  0.00172 **
## hr12                         9.2814     3.9447   2.353  0.01865 *
## hr13                        41.1417     3.9571  10.397  < 2e-16 ***
```

```
## hr14                          39.8939     3.9750  10.036  < 2e-16 ***
## hr15                          30.4940     3.9910   7.641 2.39e-14 ***
## hr16                          35.9445     3.9949   8.998  < 2e-16 ***
## hr17                          82.3786     3.9883  20.655  < 2e-16 ***
## hr18                         200.1249     3.9638  50.488  < 2e-16 ***
## hr19                         173.2989     3.9561  43.806  < 2e-16 ***
## hr20                          90.1138     3.9400  22.872  < 2e-16 ***
## hr21                          29.4071     3.9362   7.471 8.74e-14 ***
## hr22                          -8.5883     3.9332  -2.184  0.02902 *
## hr23                         -37.0194     3.9344  -9.409  < 2e-16 ***
## workingday                     1.2696     1.7845   0.711  0.47681
## temp                         157.2094    10.2612  15.321  < 2e-16 ***
## weathersitcloudy/misty       -12.8903     1.9643  -6.562 5.60e-11 ***
## weathersitlight rain/snow    -66.4944     2.9652 -22.425  < 2e-16 ***
## weathersitheavy rain/snow   -109.7446    76.6674  -1.431  0.15234
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 76.5 on 8605 degrees of freedom
## Multiple R-squared:  0.6745, Adjusted R-squared:  0.6731
## F-statistic: 457.3 on 39 and 8605 DF,  p-value: < 2.2e-16
```

```
sum((predict(mod.lm) - predict(mod.lm2))^2)
```

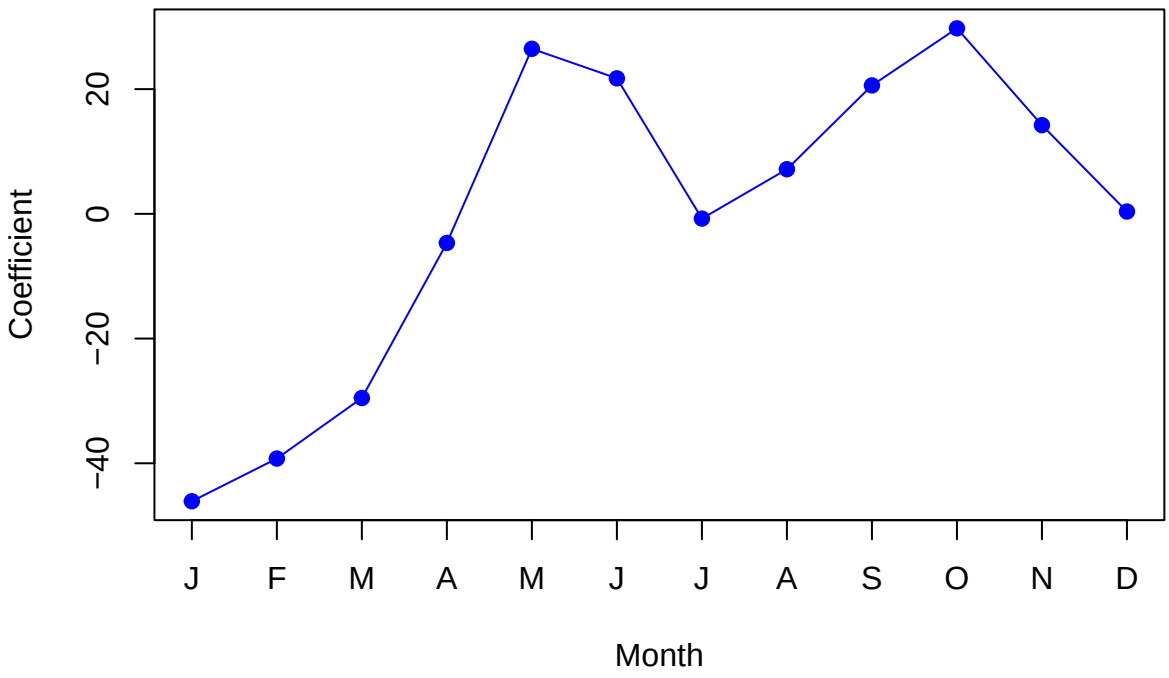
```
## [1] 1.586608e-18
```

```
all.equal(predict(mod.lm), predict(mod.lm2))
```

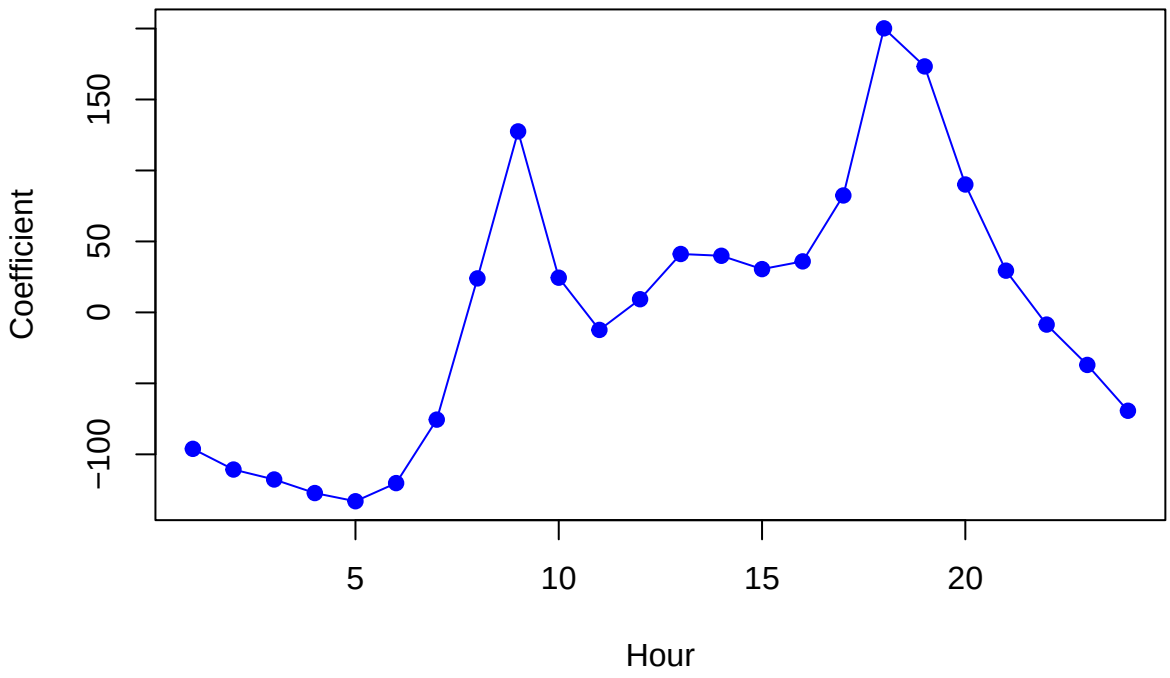
```
## [1] TRUE
```

```
coef.months <- c(coef(mod.lm2)[2:12], -sum(coef(mod.lm2)[2:12]))

plot(coef.months , xlab = "Month", ylab = "Coefficient", xaxt = "n", col = "blue", pch = 19, type = "o"
axis(side = 1, at = 1:12, labels = c("J", "F", "M", "A", "M", "J", "J", "A", "S", "O", "N", "D"))
```




```
coef.hours <- c(coef(mod.lm2)[13:35], -sum(coef(mod.lm2)[13:35]))
plot(coef.hours , xlab = "Hour", ylab = "Coefficient", col = "blue", pch = 19, type = "o")
```

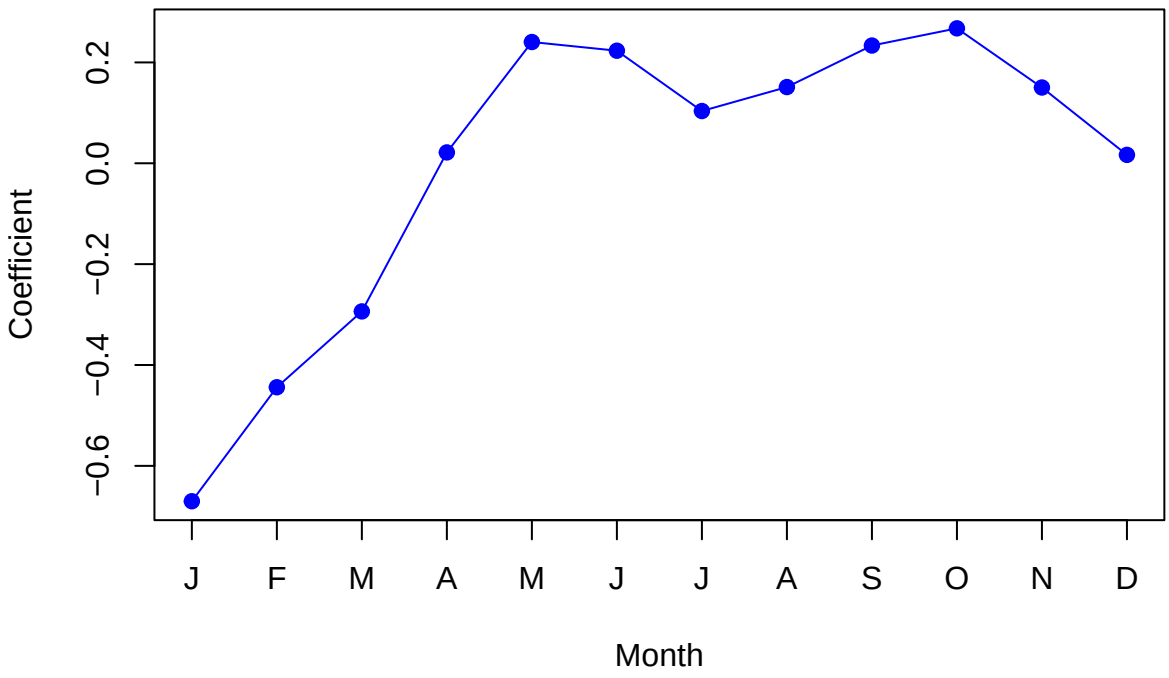



```
mod.pois <- glm(
bikers ~ mnth + hr + workingday + temp + weathersit, data = Bikeshare , family = poisson
)
summary(mod.pois)
```

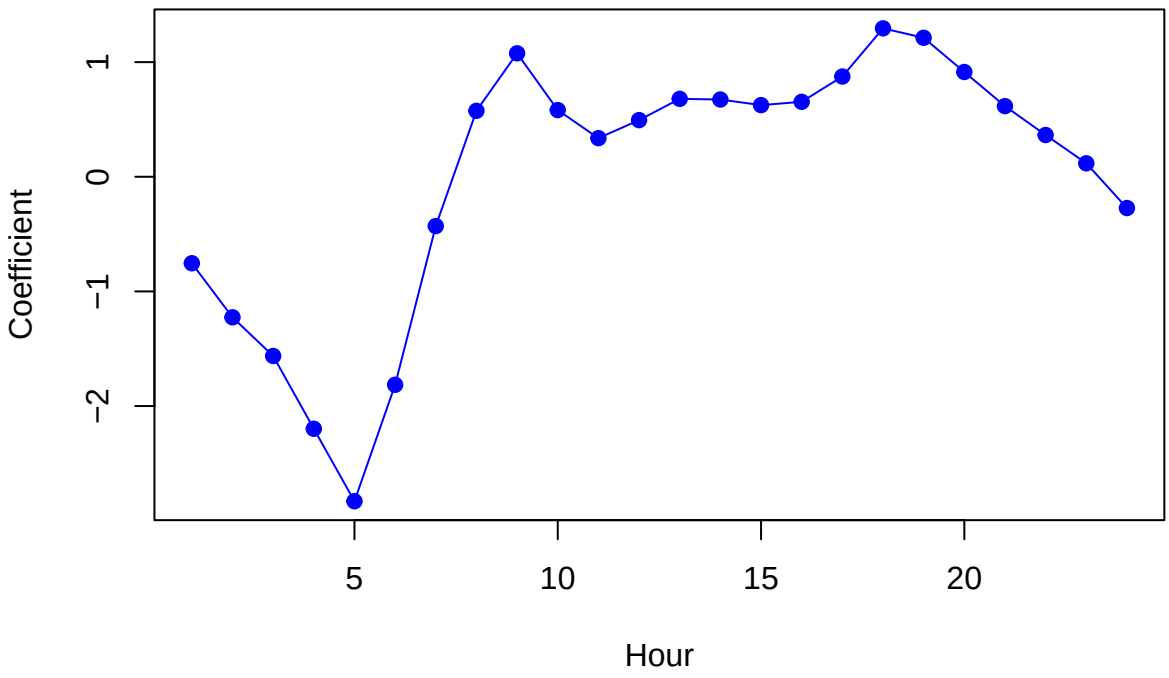
```
##
## Call:
## glm(formula = bikers ~ mnth + hr + workingday + temp + weathersit,
##     family = poisson, data = Bikeshare)
##
## Coefficients:
##                          Estimate Std. Error  z value Pr(>|z|)
## (Intercept)              4.118245   0.006021  683.964  < 2e-16 ***
## mnth1                   -0.670170   0.005907 -113.445  < 2e-16 ***
## mnth2                   -0.444124   0.004860  -91.379  < 2e-16 ***
## mnth3                   -0.293733   0.004144  -70.886  < 2e-16 ***
## mnth4                    0.021523   0.003125    6.888 5.66e-12 ***
## mnth5                    0.240471   0.002916   82.462  < 2e-16 ***
## mnth6                    0.223235   0.003554   62.818  < 2e-16 ***
## mnth7                    0.103617   0.004125   25.121  < 2e-16 ***
## mnth8                    0.151171   0.003662   41.281  < 2e-16 ***
## mnth9                    0.233493   0.003102   75.281  < 2e-16 ***
## mnth10                   0.267573   0.002785   96.091  < 2e-16 ***
## mnth11                   0.150264   0.003180   47.248  < 2e-16 ***
## hr1                     -0.754386   0.007879  -95.744  < 2e-16 ***
```

```
## hr2                        -1.225979   0.009953 -123.173  < 2e-16 ***
## hr3                        -1.563147   0.011869 -131.702  < 2e-16 ***
## hr4                        -2.198304   0.016424 -133.846  < 2e-16 ***
## hr5                        -2.830484   0.022538 -125.586  < 2e-16 ***
## hr6                        -1.814657   0.013464 -134.775  < 2e-16 ***
## hr7                        -0.429888   0.006896  -62.341  < 2e-16 ***
## hr8                         0.575181   0.004406  130.544  < 2e-16 ***
## hr9                         1.076927   0.003563  302.220  < 2e-16 ***
## hr10                        0.581769   0.004286  135.727  < 2e-16 ***
## hr11                        0.336852   0.004720   71.372  < 2e-16 ***
## hr12                        0.494121   0.004392  112.494  < 2e-16 ***
## hr13                        0.679642   0.004069  167.040  < 2e-16 ***
## hr14                        0.673565   0.004089  164.722  < 2e-16 ***
## hr15                        0.624910   0.004178  149.570  < 2e-16 ***
## hr16                        0.653763   0.004132  158.205  < 2e-16 ***
## hr17                        0.874301   0.003784  231.040  < 2e-16 ***
## hr18                        1.294635   0.003254  397.848  < 2e-16 ***
## hr19                        1.212281   0.003321  365.084  < 2e-16 ***
## hr20                        0.914022   0.003700  247.065  < 2e-16 ***
## hr21                        0.616201   0.004191  147.045  < 2e-16 ***
## hr22                        0.364181   0.004659   78.173  < 2e-16 ***
## hr23                        0.117493   0.005225   22.488  < 2e-16 ***
## workingday                  0.014665   0.001955    7.502 6.27e-14 ***
## temp                        0.785292   0.011475   68.434  < 2e-16 ***
## weathersitcloudy/misty     -0.075231   0.002179  -34.528  < 2e-16 ***
## weathersitlight rain/snow -0.575800   0.004058 -141.905  < 2e-16 ***
## weathersitheavy rain/snow -0.926287   0.166782   -5.554 2.79e-08 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for poisson family taken to be 1)
##
##     Null deviance: 1052921  on 8644  degrees of freedom
## Residual deviance:  228041  on 8605  degrees of freedom
## AIC: 281159
##
## Number of Fisher Scoring iterations: 5
```

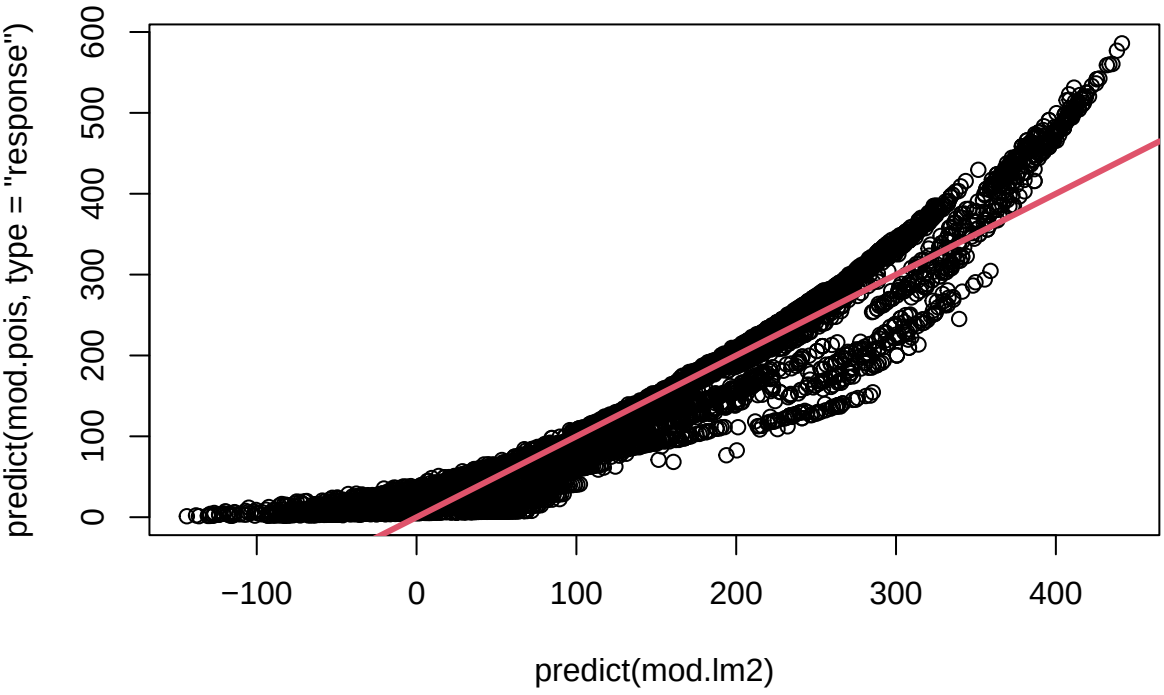
```r
coef.mnth <- c(coef(mod.pois)[2:12], -sum(coef(mod.pois)[2:12]))
plot(coef.mnth , xlab = "Month", ylab = "Coefficient", xaxt = "n", col = "blue", pch = 19, type = "o")
axis(side = 1, at = 1:12, labels = c("J", "F", "M", "A", "M", "J", "J", "A", "S", "O", "N", "D"))
```




```
coef.hours <- c(coef(mod.pois)[13:35], -sum(coef(mod.pois)[13:35]))
plot(coef.hours , xlab = "Hour", ylab = "Coefficient", col = "blue", pch = 19, type = "o")
```




```
plot(predict(mod.lm2), predict(mod.pois , type = "response"))
abline(0, 1, col = 2, lwd = 3)
```


predict(mod.pois, type = "response")
0
100
200
300
400
500
600
−100
0
100
200
300
400
predict(mod.lm2)