

A5

A00829752

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```
library(ISLR)
library(tidyverse)
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr      1.1.3      v readr      2.1.4
## v forcats    1.0.0      v stringr    1.5.0
## v ggplot2     3.4.3      v tibble     3.2.1
## v lubridate  1.9.3      v tidyr      1.3.0
## v purrr       1.0.2
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors

dataSP <- Weekly
```

```
head(Weekly)
```

	Year	Lag1	Lag2	Lag3	Lag4	Lag5	Volume	Today	Direction
## 1	1990	0.816	1.572	-3.936	-0.229	-3.484	0.1549760	-0.270	Down
## 2	1990	-0.270	0.816	1.572	-3.936	-0.229	0.1485740	-2.576	Down
## 3	1990	-2.576	-0.270	0.816	1.572	-3.936	0.1598375	3.514	Up
## 4	1990	3.514	-2.576	-0.270	0.816	1.572	0.1616300	0.712	Up
## 5	1990	0.712	3.514	-2.576	-0.270	0.816	0.1537280	1.178	Up
## 6	1990	1.178	0.712	3.514	-2.576	-0.270	0.1544440	-1.372	Down

```
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	Today
## Min.	:-18.1950	Min. :-18.1950	Min. :0.08747	Min. :-18.1950
## 1st Qu.:	-1.1580	1st Qu.: -1.1660	1st Qu.:0.33202	1st Qu.: -1.1540
## Median :	0.2380	Median : 0.2340	Median :1.00268	Median : 0.2410
## Mean :	0.1458	Mean : 0.1399	Mean :1.57462	Mean : 0.1499
## 3rd Qu.:	1.4090	3rd Qu.: 1.4050	3rd Qu.:2.05373	3rd Qu.: 1.4050
## Max. :	12.0260	Max. : 12.0260	Max. :9.32821	Max. : 12.0260

```
## Direction
## Down:484
```

```
## Up :605
##
##
##
##
```

```
cor(Weekly[, -9])
```

```
##           Year      Lag1      Lag2      Lag3      Lag4
## Year    1.00000000 -0.032289274 -0.03339001 -0.03000649 -0.031127923
## Lag1    -0.03228927  1.000000000 -0.07485305  0.05863568 -0.071273876
## Lag2    -0.03339001 -0.074853051  1.00000000 -0.07572091  0.058381535
## Lag3    -0.03000649  0.058635682 -0.07572091  1.00000000 -0.075395865
## Lag4    -0.03112792 -0.071273876  0.05838153 -0.07539587  1.000000000
## Lag5    -0.03051910 -0.008183096 -0.07249948  0.06065717 -0.075675027
## Volume   0.84194162 -0.064951313 -0.08551314 -0.06928771 -0.061074617
## Today   -0.03245989 -0.075031842  0.05916672 -0.07124364 -0.007825873
##           Lag5      Volume      Today
## Year   -0.030519101  0.84194162 -0.032459894
## Lag1   -0.008183096 -0.06495131 -0.075031842
## Lag2   -0.072499482 -0.08551314  0.059166717
## Lag3    0.060657175 -0.06928771 -0.071243639
## Lag4   -0.075675027 -0.06107462 -0.007825873
## Lag5    1.000000000 -0.05851741  0.011012698
## Volume -0.058517414  1.000000000 -0.033077783
## Today   0.011012698 -0.03307778  1.000000000
```

```
#Generacion del modelo
```

```
modelo.log.m <- glm(Direction ~ .-Today, data = Weekly, family = binomial)
summary(modelo.log.m)
```

```
##
## Call:
## glm(formula = Direction ~ . - Today, family = binomial, data = Weekly)
##
## Coefficients:
##           Estimate Std. Error z value Pr(>|z|)
## (Intercept) 17.225822  37.890522  0.455  0.6494
## Year        -0.008500  0.018991 -0.448  0.6545
## Lag1        -0.040688  0.026447 -1.538  0.1239
## Lag2         0.059449  0.026970  2.204  0.0275 *
## Lag3        -0.015478  0.026703 -0.580  0.5622
## Lag4        -0.027316  0.026485 -1.031  0.3024
## Lag5        -0.014022  0.026409 -0.531  0.5955
## Volume       0.003256  0.068836  0.047  0.9623
## ---
## 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: 1486.2 on 1081 degrees of freedom
## AIC: 1502.2
##
## Number of Fisher Scoring iterations: 4
```

```

confint(object = modelo.log.m, level = 0.95)

## Waiting for profiling to be done...

##              2.5 %      97.5 %
## (Intercept) -56.985558236  91.66680901
## Year        -0.045809580   0.02869546
## Lag1         -0.092972584   0.01093101
## Lag2          0.007001418   0.11291264
## Lag3         -0.068140141   0.03671410
## Lag4         -0.079519582   0.02453326
## Lag5         -0.066090145   0.03762099
## Volume       -0.131576309   0.13884038

#Division de la base de datos
train_data <- Weekly[Weekly$Year < 2009, ]
test_data  <- Weekly[Weekly$Year %in% c(2009, 2010), ]

model_significant <- glm(Direction ~ Lag1 + Lag2 + Lag4, data = train_data, family = binomial)
summary(model_significant)

##
## Call:
## glm(formula = Direction ~ Lag1 + Lag2 + Lag4, family = binomial,
##      data = train_data)
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  0.21393    0.06471   3.306 0.000946 ***
## Lag1        -0.05551    0.02891  -1.920 0.054848 .
## Lag2         0.05530    0.02913   1.898 0.057676 .
## Lag4        -0.02094    0.02866  -0.731 0.464886
## ---
## 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: 1346.4  on 981  degrees of freedom
## AIC: 1354.4
##
## Number of Fisher Scoring iterations: 4

library(caret)

## Loading required package: lattice

##
## Attaching package: 'caret'

## The following object is masked from 'package:purrr':
##
##      lift

predictions <- predict(model_significant, newdata = test_data, type = "response")
predictions <- factor(ifelse(predictions > 0.5, "Up", "Down"), levels = c("Down", "Up"))

```

```
conf_matrix <- confusionMatrix(predictions, test_data$Direction)
conf_matrix
```

```
## Confusion Matrix and Statistics
##
##           Reference
## Prediction Down Up
##      Down    9  7
##      Up     34 54
##
##           Accuracy : 0.6058
##           95% CI : (0.5051, 0.7002)
##      No Information Rate : 0.5865
##      P-Value [Acc > NIR] : 0.3847
##
##           Kappa : 0.1042
##
##  Mcnemar's Test P-Value : 4.896e-05
##
##           Sensitivity : 0.20930
##           Specificity : 0.88525
##      Pos Pred Value : 0.56250
##      Neg Pred Value : 0.61364
##           Prevalence : 0.41346
##      Detection Rate : 0.08654
##      Detection Prevalence : 0.15385
##      Balanced Accuracy : 0.54727
##
##      'Positive' Class : Down
##
```

```
observed <- table(predictions, test_data$Direction)
chi_squared_test <- chisq.test(observed)
print(chi_squared_test)
```

```
##
##  Pearson's Chi-squared test with Yates' continuity correction
##
## data:  observed
## X-squared = 1.0818, df = 1, p-value = 0.2983
```

```
coeficientes <- coef(model_significant)
print(coeficientes)
```

```
## (Intercept)      Lag1      Lag2      Lag4
##  0.21393356 -0.05550915  0.05529514 -0.02094275
```

El modelo no nos brinda los mejores resultados por lo que seria recomendado continuar haciendo ajustes.