

# Regresión Logística 2021 Reto

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2023-10-19

## Regresión Logística Multinomial

### Librerías necesarias y lectura de datos

```
library(nnet)
library(tidyverse)

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

library(caret)

## Loading required package: lattice
##
## Attaching package: 'caret'
##
## The following object is masked from 'package:purrr':
##
##   lift

library(MASS)

##
## Attaching package: 'MASS'
##
## The following object is masked from 'package:dplyr':
##
##   select
```

```
data <- read.csv("data_2021_RegLog.csv")
```

```
head(data)
```

```
##      PRS RH      SR TOUT  WSR WDR O3Concentracion
## 1 698.9 43 0.059 20.83  8.7 109                3
## 2 706.8 29 0.881 36.61 12.9  56                3
## 3 704.4 16 0.946 34.57 10.0  64                3
## 4 706.8 49 0.336 30.92 10.0 136                3
## 5 697.5 37 0.041 28.39 14.1 102                3
## 6 697.8 46 0.044 26.42 11.7 107                3
```

## Dividir datos en entrenamiento y prueba

```
set.seed(123)
```

```
trainIndex <- createDataPartition(data$O3Concentracion, p = 0.8, list = FALSE)
```

```
train_data <- data[trainIndex, ]
test_data <- data[-trainIndex, ]
```

## Ajustar el modelo de regresión y Resumen

```
set.seed(123)
```

```
# Ajustar el modelo
```

```
model <- multinom(O3Concentracion ~ ., data = train_data)
```

```
## # weights:  24 (14 variable)
## initial value 23031.308020
## iter  10 value 5959.477674
## iter  20 value 4009.868276
## iter  30 value 3791.876428
## iter  40 value 3771.061101
## final value 3734.906691
## converged
```

```
summary(model)
```

```
## Call:
## multinom(formula = O3Concentracion ~ ., data = train_data)
##
## Coefficients:
##      (Intercept)          PRS          RH          SR          TOUT          WSR
## 2    -7.082130    0.006137756 -0.04822663  2.904467  0.07008665  0.01461195
## 3     6.148726   -0.024837003 -0.02497572  3.686489  0.23219355 -0.04726438
##
##              WDR
## 2   -0.004860281
## 3   -0.001101421
##
## Std. Errors:
```

```
##      (Intercept)          PRS          RH          SR          TOUT
WSR
## 2 1.717799e-03 0.0002891926 0.002106992 0.114601215 0.00675806
0.007843537
## 3 4.953108e-05 0.0019725671 0.009904873 0.002953627 0.03917913
0.033425292
##              WDR
## 2 0.0004807038
## 3 0.0016679317
##
## Residual Deviance: 7469.813
## AIC: 7497.813
```

## Precisión del modelo

*# Predecir en el conjunto de prueba*

```
predicted_probs <- predict(model, newdata = test_data, type = "probs")
predicted_class <- predict(model, newdata = test_data, type = "class")
```

*# Convertir tanto las predicciones como la variable real a factores con niveles 1, 2, 3 y 4.*

```
predicted_class <- factor(predicted_class, levels = c(1, 2, 3))
true_class <- factor(test_data$O3Concentracion, levels = c(1, 2, 3))
```

*# Ahora crea la matriz de confusión*

```
confusionMatrix(predicted_class, true_class)
```

```
## Confusion Matrix and Statistics
```

```
##
```

```
##           Reference
## Prediction    1    2    3
##           1 4870  279  15
##           2   34   40   3
##           3    0    0    0
```

```
##
```

```
## Overall Statistics
```

```
##
```

```
##              Accuracy : 0.9368
##              95% CI : (0.9299, 0.9433)
##      No Information Rate : 0.9357
##      P-Value [Acc > NIR] : 0.3812
```

```
##
```

```
##              Kappa : 0.1814
```

```
##
```

```
## McNemar's Test P-Value : <2e-16
```

```
##
```

```
## Statistics by Class:
```

```
##
```

```
##              Class: 1 Class: 2 Class: 3
## Sensitivity      0.9931 0.125392 0.000000
```

```
## Specificity          0.1276 0.992483 1.000000
## Pos Pred Value      0.9431 0.519481      NaN
## Neg Pred Value      0.5584 0.945972 0.996566
## Prevalence          0.9357 0.060866 0.003434
## Detection Rate      0.9292 0.007632 0.000000
## Detection Prevalence 0.9853 0.014692 0.000000
## Balanced Accuracy    0.5603 0.558937 0.500000
```

### Valor p de coeficientes

```
coefficients <- coef(model)
std_errors <- summary(model)$standard.errors
t_statistics <- coefficients / std_errors
df <- nrow(train_data) - length(coefficients)
p_values <- 2 * (1 - pt(abs(t_statistics), df))
p_values
```

	(Intercept)	PRS	RH	SR	TOUT	WSR	WDR
## 2	0	0	0.00000000	0	0.000000e+00	0.06248631	0.00000000
## 3	0	0	0.01169095	0	3.14379e-09	0.15736783	0.5090357

### Coeficientes del modelo

```
coef(model)
```

	(Intercept)	PRS	RH	SR	TOUT	WSR
## 2	-7.082130	0.006137756	-0.04822663	2.904467	0.07008665	0.01461195
## 3	6.148726	-0.024837003	-0.02497572	3.686489	0.23219355	-0.04726438

```
## WDR
## 2 -0.004860281
## 3 -0.001101421
```

### Significancia de variables mediante la prueba de Wald

```
library(nnet)
library(car)

## Loading required package: carData

##
## Attaching package: 'car'

## The following object is masked from 'package:dplyr':
##
## recode

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

Anova(model, type="II")

## Analysis of Deviance Table (Type II tests)
##
```

```
## Response: O3Concentracion
##      LR Chisq Df Pr(>Chisq)
## PRS      9.95  2  0.006926 **
## RH     538.31  2 < 2.2e-16 ***
## SR     661.05  2 < 2.2e-16 ***
## TOUT    157.04  2 < 2.2e-16 ***
## WSR       5.56  2  0.062006 .
## WDR     103.88  2 < 2.2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
library(car)
Anova(model, test="Wald")
```

```
## Analysis of Deviance Table (Type II tests)
##
## Response: O3Concentracion
##      LR Chisq Df Pr(>Chisq)
## PRS      9.95  2  0.006926 **
## RH     538.31  2 < 2.2e-16 ***
## SR     661.05  2 < 2.2e-16 ***
## TOUT    157.04  2 < 2.2e-16 ***
## WSR       5.56  2  0.062006 .
## WDR     103.88  2 < 2.2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```