RML\_042.R

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2020-04-02

# Rozdział 4 - regresja logistyczna  
  
rm(list=ls())  
  
# Dane giełdowe indeksu S&P zmian dziennych  
  
library(ISLR)

## Warning: package 'ISLR' was built under R version 3.6.3

names(Smarket)

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

dim(Smarket)

## [1] 1250 9

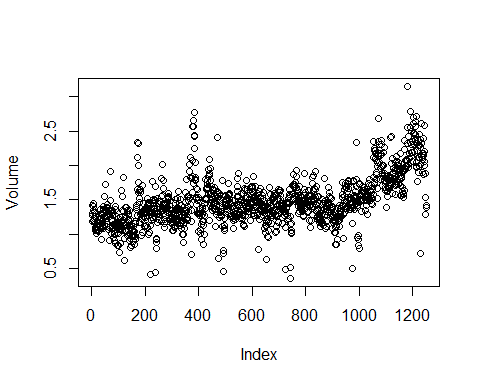
summary(Smarket)

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

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

attach(Smarket)  
plot(Volume)



# Regresja logistyczna  
  
glm.fit=glm(Direction~Lag1+Lag2+Lag3+Lag4+Lag5+Volume,data=Smarket,family=binomial)  
summary(glm.fit)

##   
## Call:  
## glm(formula = Direction ~ Lag1 + Lag2 + Lag3 + Lag4 + Lag5 +   
## Volume, family = binomial, data = Smarket)  
##   
## Deviance Residuals:   
## Min 1Q Median 3Q Max   
## -1.446 -1.203 1.065 1.145 1.326   
##   
## 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.fit)

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

summary(glm.fit)$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.fit)$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.fit,type="response")  
glm.probs[1:10]

## 1 2 3 4 5 6 7   
## 0.5070841 0.4814679 0.4811388 0.5152224 0.5107812 0.5069565 0.4926509   
## 8 9 10   
## 0.5092292 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.fit=glm(Direction~Lag1+Lag2+Lag3+Lag4+Lag5+Volume,data=Smarket,family=binomial,subset=train)  
glm.probs=predict(glm.fit,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.fit=glm(Direction~Lag1+Lag2,data=Smarket,family=binomial,subset=train)  
glm.probs=predict(glm.fit,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.fit,newdata=data.frame(Lag1=c(1.2,1.5),Lag2=c(1.1,-0.8)),type="response")

## 1 2   
## 0.4791462 0.4960939

# Liniowa analiza dyskryminacyjna  
  
library(MASS)  
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

mean(lda.class==Direction.2005)

## [1] 0.5595238

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   
## 0.4901792 0.4792185 0.4668185 0.4740011 0.4927877 0.4938562 0.4951016   
## 1006 1007 1008 1009 1010 1011 1012   
## 0.4872861 0.4907013 0.4844026 0.4906963 0.5119988 0.4895152 0.4706761   
## 1013 1014 1015 1016 1017 1018   
## 0.4744593 0.4799583 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   
## [15] Up Up Up Down Up Up   
## Levels: Down Up

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

## [1] 0

# Kwadratowa analiza dyskryminacyjna  
  
qda.fit=qda(Direction~Lag1+Lag2,data=Smarket,subset=train)  
qda.fit

## 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

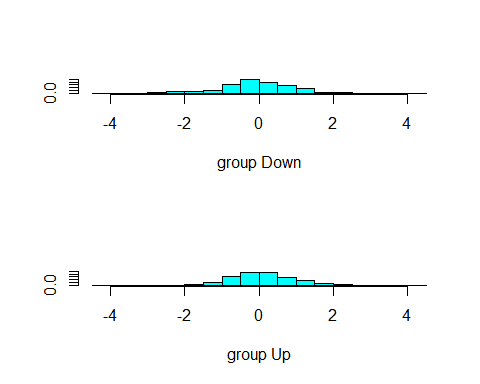
mean(qda.class==Direction.2005)

## [1] 0.5992063

#definiowanie sprawdzianu krzyżowego  
library(caret)

## Loading required package: lattice

## Loading required package: ggplot2



library(e1071)  
train\_control <- trainControl(method="cv", number=10)  
  
#regresja logistyczna  
model <- train(Direction~Lag1+Lag2+Lag3+Lag4+Lag5+Volume,data=Smarket, trControl=train\_control, method="glm", family="binomial")  
  
#sprawdzenie modelu i porównanie z wcześniejszym  
summary(model)

##   
## Call:  
## NULL  
##   
## Deviance Residuals:   
## Min 1Q Median 3Q Max   
## -1.446 -1.203 1.065 1.145 1.326   
##   
## 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

summary(glm.fit)

##   
## Call:  
## glm(formula = Direction ~ Lag1 + Lag2, family = binomial, data = Smarket,   
## subset = train)  
##   
## Deviance Residuals:   
## Min 1Q Median 3Q Max   
## -1.345 -1.188 1.074 1.164 1.326   
##   
## Coefficients:  
## Estimate Std. Error z value Pr(>|z|)  
## (Intercept) 0.03222 0.06338 0.508 0.611  
## Lag1 -0.05562 0.05171 -1.076 0.282  
## Lag2 -0.04449 0.05166 -0.861 0.389  
##   
## (Dispersion parameter for binomial family taken to be 1)  
##   
## Null deviance: 1383.3 on 997 degrees of freedom  
## Residual deviance: 1381.4 on 995 degrees of freedom  
## AIC: 1387.4  
##   
## Number of Fisher Scoring iterations: 3

#Dokłdność modelu  
print(model)

## Generalized Linear Model   
##   
## 1250 samples  
## 6 predictor  
## 2 classes: 'Down', 'Up'   
##   
## No pre-processing  
## Resampling: Cross-Validated (10 fold)   
## Summary of sample sizes: 1126, 1125, 1124, 1125, 1125, 1125, ...   
## Resampling results:  
##   
## Accuracy Kappa   
## 0.5055362 -0.008841778