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數據分析與應用 作業 4:鋼鐵瑕疵影像分類實作
#載入套件
install.packages(c("imager","caret","EBImage"))
library(imager)
library(caret)
library(EBImage)
#load train data .jpg
root = 'C:/Users/MCUT/Desktop/Steel/train/'
filenames = list.files(path = root, pattern = ".jpg", recursive = TRUE)
feature = {}
label = \{\}
for (i in filenames) {
  img = readImage(paste(root, i, sep = ""))
  label = c(label, strsplit(i, "[/]")[[1]][1])
  feature = rbind(feature, computeFeatures.basic(x=matrix(1, dim(img)[1],
dim(img)[2]), ref=img))
}
trainset = data.frame(feature, label)
#load test data .jpg
root = 'C:/Users/MCUT/Desktop/Steel/test/'
filenames = list.files(path = root, pattern = ".jpg", recursive = TRUE)
feature = {}
label = \{\}
for (i in filenames) {
  img = readImage(paste(root, i, sep = ""))
  label = c(label, strsplit(i, "[/]")[[1]][1])
  feature = rbind(feature, computeFeatures.basic(x=matrix(1, dim(img)[1],
dim(img)[2]), ref=img))
testset = data.frame(feature, label)
# train model
#set.seed(7)
control fit <- trainControl(method='cv', number=5,
                                  classProbs=TRUE,
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summaryFunction=multiClassSummary,
selectionFunction = 'best')
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glmnet_fit<- caret::train(label~., data=trainset,

method='glmnet', metric='Accuracy', trControl=control_fit, verbose = FALSE)

#predict

pred_res<-predict(glmnet_fit, newdata=testset)</pre>

#confusionMatrix

confusionMatrix1<-confusionMatrix(pred_res,

factor(testset\$label),
dnn = c('Prediction', 'Reference'),
mode = 'everything')

> confusionMatrix1

Confusion Matrix and Statistics

Reference

Prediction	nail_jpg	scratch_jpg	smear_jpg
nail_jpg	23	6	0
scratch_jpg	3	8	0
smear ipg	2	0	21

Overall Statistics

Accuracy: 0.8254 95% CI: (0.709, 0.9095)

No Information Rate : 0.4444 P-Value [Acc > NIR] : 5.56e-10

Карра : 0.725

Mcnemar's Test P-Value : NA

Statistics by class:

	Class: nail_jpg Class:	scratch_jpg Class:	smear_jpg
Sensitivity	0.8214	0.5714	1.0000
Specificity	0.8286	0.9388	0.9524
Pos Pred Value	0.7931	0.7273	0.9130
Neg Pred Value	0.8529	0.8846	1.0000
Precision	0.7931	0.7273	0.9130
Recall	0.8214	0.5714	1.0000
F1	0.8070	0.6400	0.9545
Prevalence	0.4444	0.2222	0.3333
Detection Rate	0.3651	0.1270	0.3333
Detection Prevalence	0.4603	0.1746	0.3651
Balanced Accuracy	0.8250	0.7551	0.9762

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# preProcess
preProValues<-preProcess(trainset, method = c('center', 'scale'))</pre>
traindata<-predict(preProValues,trainset)
testdata<-predict(preProValues,testset)
#train glmnet model
control fit <- trainControl(method='cv', number=5,
                                  classProbs=TRUE,
                                  summaryFunction=multiClassSummary,
                                  selectionFunction = 'best')
glmnet fit<- caret::train(label~., data=traindata,
                                   method='glmnet',
                                   metric='Accuracy',
                                   trControl=control fit,
                                   verbose = FALSE)
#predict
pred_res<-predict(glmnet_fit, newdata=testdata)</pre>
confusionMatrix1<-confusionMatrix(pred_res,
                                         factor(testdata$label),
                                          dnn = c('Prediction', 'Reference'),
                                          mode = 'everything')
      Confusion Matrix and Statistics
                   Reference
      Prediction
                  nail_jpg scratch_jpg smear_jpg
       nail_jpg
                           3
                                                 0
        scratch_jpg
        smear_jpg
                           2
                                       0
                                                21
      Overall Statistics
                     Accuracy: 0.8095
                       95% CI: (0.6909, 0.8975)
          No Information Rate: 0.4444
          P-Value [Acc > NIR] : 3.078e-09
                        карра: 0.6983
      Mcnemar's Test P-Value : NA
      Statistics by Class:
                           class: nail_jpg Class: scratch_jpg Class: smear_jpg
      Sensitivity
                                                       0.5000
                                    0.8000
      Specificity
                                                       0.9388
                                                                        0.9524
      Pos Pred Value
                                    0.7667
                                                       0.7000
                                                                        0.9130
      Neg Pred Value
                                    0.8485
                                                       0.8679
                                                                        1.0000
      Précision
                                    0.7667
                                                       0.7000
                                                                        0.9130
                                                       0.5000
                                                                        1.0000
      Recall.
                                    0.8214
                                   0.7931
                                                       0.5833
                                                                        0.9545
      Prevalence
                                   0.4444
                                                       0.2222
                                                                        0.3333
                                                       0.1111
      Detection Rate
                                   0.3651
                                                                        0.3333
     Detection Prevalence
```

0.4762

0.8107

Balanced Accuracy

0.1587

0.7194

0.3651

0.9762