

Assignment_2

Dengcheng Chen

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```
library(readr)
Ubank <- read_delim(file = 'UniversalBank.csv',delim=',')

## Rows: 5000 Columns: 14
## -- Column specification -----
## Delimiter: ","
## dbl (14): ID, Age, Experience, Income, ZIP Code, Family, CCAvg, Education,
## M...
## i Use `spec()` to retrieve the full column specification for this data.
## i Specify the column types or set `show_col_types = FALSE` to quiet this message.

library(reshape)
Ubank <- rename(Ubank,c(`Personal Loan` = 'PL','Securities Account'='SA','CD
Account'='CDA'))
names(Ubank)

## [1] "ID"          "Age"          "Experience"   "Income"       "ZIP Code"
## [6] "Family"      "CCAvg"        "Education"    "Mortgage"     "PL"
## [11] "SA"          "CDA"          "Online"       "CreditCard"

summary(Ubank)

##      ID          Age      Experience      Income      ZIP Cod
## Min.   : 1    Min.   :23.00   Min.   : -3.0   Min.   : 8.00   Min.   : 9
307
## 1st Qu.:1251   1st Qu.:35.00   1st Qu.:10.0   1st Qu.: 39.00   1st Qu.:91
911
## Median :2500   Median :45.00   Median :20.0   Median : 64.00   Median :93
437
## Mean    :2500   Mean    :45.34   Mean    :20.1   Mean    : 73.77   Mean    :93
153
## 3rd Qu.:3750   3rd Qu.:55.00   3rd Qu.:30.0   3rd Qu.: 98.00   3rd Qu.:94
608
## Max.    :5000   Max.    :67.00   Max.    :43.0   Max.    :224.00   Max.    :96
651
##      Family      CCAvg      Education      Mortgage
## Min.   :1.000    Min.   : 0.000    Min.   :1.000    Min.   : 0.0
## 1st Qu.:1.000    1st Qu.: 0.700    1st Qu.:1.000    1st Qu.: 0.0
## Median :2.000    Median : 1.500    Median :2.000    Median : 0.0
```

```
## Mean :2.396 Mean : 1.938 Mean :1.881 Mean : 56.5
## 3rd Qu.:3.000 3rd Qu.: 2.500 3rd Qu.:3.000 3rd Qu.:101.0
## Max. :4.000 Max. :10.000 Max. :3.000 Max. :635.0
## PL SA CDA Online
## Min. :0.000 Min. :0.0000 Min. :0.0000 Min. :0.0000
## 1st Qu.:0.000 1st Qu.:0.0000 1st Qu.:0.0000 1st Qu.:0.0000
## Median :0.000 Median :0.0000 Median :0.0000 Median :1.0000
## Mean :0.096 Mean :0.1044 Mean :0.0604 Mean :0.5968
## 3rd Qu.:0.000 3rd Qu.:0.0000 3rd Qu.:0.0000 3rd Qu.:1.0000
## Max. :1.000 Max. :1.0000 Max. :1.0000 Max. :1.0000
## CreditCard
## Min. :0.000
## 1st Qu.:0.000
## Median :0.000
## Mean :0.294
## 3rd Qu.:1.000
## Max. :1.000
```

```
Ubank$ID <- NULL
Ubank$`ZIP Code` <- NULL
```

```
Ubank$Education = as.factor(Ubank$Education)
Ubank$PL = as.factor(Ubank$PL)
summary(Ubank)
```

```
## Age Experience Income Family
## Min. :23.00 Min. :-3.0 Min. : 8.00 Min. :1.000
## 1st Qu.:35.00 1st Qu.:10.0 1st Qu.: 39.00 1st Qu.:1.000
## Median :45.00 Median :20.0 Median : 64.00 Median :2.000
## Mean :45.34 Mean :20.1 Mean : 73.77 Mean :2.396
## 3rd Qu.:55.00 3rd Qu.:30.0 3rd Qu.: 98.00 3rd Qu.:3.000
## Max. :67.00 Max. :43.0 Max. :224.00 Max. :4.000
## CCAvg Education Mortgage PL SA
## Min. : 0.000 1:2096 Min. : 0.0 0:4520 Min. :0.0000
## 1st Qu.: 0.700 2:1403 1st Qu.: 0.0 1: 480 1st Qu.:0.0000
## Median : 1.500 3:1501 Median : 0.0 Median :0.0000
## Mean : 1.938 Mean : 56.5 Mean :0.1044
## 3rd Qu.: 2.500 3rd Qu.:101.0 3rd Qu.:0.0000
## Max. :10.000 Max. :635.0 Max. :1.0000
## CDA Online CreditCard
## Min. :0.0000 Min. :0.0000 Min. :0.000
## 1st Qu.:0.0000 1st Qu.:0.0000 1st Qu.:0.000
## Median :0.0000 Median :1.0000 Median :0.000
## Mean :0.0604 Mean :0.5968 Mean :0.294
## 3rd Qu.:0.0000 3rd Qu.:1.0000 3rd Qu.:1.000
## Max. :1.0000 Max. :1.0000 Max. :1.000
```

```
library(caret)
```

```
## 载入需要的程辑包: ggplot2
```

```
## 载入需要的程辑包: lattice

library(class)

##
## 载入程辑包: 'class'

## The following object is masked from 'package:reshape':
##
##      condense

dummies <- dummyVars(PL ~ ., data = Ubank)
Ubank_dummy <- as.data.frame(predict(dummies, newdata= Ubank))

## Warning in model.frame.default(Terms, newdata, na.action = na.action, xlev
=
## object$lvls): variable 'PL' is not a factor

head(Ubank_dummy)

##   Age Experience Income Family CCAvg Education.1 Education.2 Education.3
## 1  25          1     49      4   1.6          1          0          0
## 2  45         19     34      3   1.5          1          0          0
## 3  39         15     11      1   1.0          1          0          0
## 4  35          9    100      1   2.7          0          1          0
## 5  35          8     45      4   1.0          0          1          0
## 6  37         13     29      4   0.4          0          1          0
##   Mortgage SA CDA Online CreditCard
## 1         0  1  0      0          0
## 2         0  1  0      0          0
## 3         0  0  0      0          0
## 4         0  0  0      0          0
## 5         0  0  0      0          1
## 6        155  0  0      1          0

Norm_model <- preProcess(Ubank_dummy, method = c("center", "scale"))
Ubank_norm = predict(Norm_model, Ubank_dummy)
summary(Ubank_norm)

##           Age           Experience           Income           Family
##  Min.      :-1.94871   Min.      :-2.014710   Min.      :-1.4288   Min.      :-1.2167
## 1st Qu.: -0.90188   1st Qu.: -0.881116   1st Qu.: -0.7554   1st Qu.: -1.2167
##  Median :-0.02952   Median :-0.009121   Median :-0.2123   Median :-0.3454
##  Mean     : 0.00000   Mean      : 0.000000   Mean      : 0.0000   Mean      : 0.0000
## 3rd Qu.: 0.84284   3rd Qu.: 0.862874   3rd Qu.: 0.5263   3rd Qu.: 0.5259
```

```
## Max. : 1.88967 Max. : 1.996468 Max. : 3.2634 Max. : 1.3973

## CCAvg Education.1 Education.2 Education.3
## Min. :-1.1089 Min. :-0.8495 Min. :-0.6245 Min. :-0.6549
## 1st Qu.:-0.7083 1st Qu.:-0.8495 1st Qu.:-0.6245 1st Qu.:-0.6549
## Median :-0.2506 Median :-0.8495 Median :-0.6245 Median :-0.6549
## Mean : 0.0000 Mean : 0.0000 Mean : 0.0000 Mean : 0.0000
## 3rd Qu.: 0.3216 3rd Qu.: 1.1770 3rd Qu.: 1.6010 3rd Qu.: 1.5266
## Max. : 4.6131 Max. : 1.1770 Max. : 1.6010 Max. : 1.5266
## Mortgage SA CDA Online
## Min. :-0.5555 Min. :-0.3414 Min. :-0.2535 Min. :-1.2165
## 1st Qu.:-0.5555 1st Qu.:-0.3414 1st Qu.:-0.2535 1st Qu.:-1.2165
## Median :-0.5555 Median :-0.3414 Median :-0.2535 Median : 0.8219
## Mean : 0.0000 Mean : 0.0000 Mean : 0.0000 Mean : 0.0000
## 3rd Qu.: 0.4375 3rd Qu.:-0.3414 3rd Qu.:-0.2535 3rd Qu.: 0.8219
## Max. : 5.6875 Max. : 2.9286 Max. : 3.9438 Max. : 0.8219
## CreditCard
## Min. :-0.6452
## 1st Qu.:-0.6452
## Median :-0.6452
## Mean : 0.0000
## 3rd Qu.: 1.5495
## Max. : 1.5495
```

```
Ubank_norm$PL=Ubank$PL
```

```
Train_Index = createDataPartition(Ubank$PL,p=0.6, list=FALSE)
Train.df = Ubank_norm[Train_Index,]
Validation.df = Ubank_norm[-Train_Index,]
```

Q1 Age = 40, Experience = 10, Income = 84, Family = 2, CCAvg = 2, Education_1 = 0, Education_2 = 1, Education_3 = 0, Mortgage = 0, Securities Account = 0, CD Account = 0, Online = 1, and CreditCard = 1. Perform a k-NN classification with all predictors except ID and ZIP code using k = 1. Remember to transform categorical predictors with more than two categories into dummy variables first. Specify the success class as 1 (loan acceptance), and use the default cutoff value of 0.5. How would this customer be classified?

```
To_Predict=data.frame(Age=40, Experience = 10, Income = 84, Family = 2, CCAvg
= 2,
                        Education.1 = 0, Education.2 = 1, Education.3 = 0, Mort
gage = 0,
                        SA = 0, CDA = 0, Online = 1,CreditCard = 1)

print(To_Predict)

## Age Experience Income Family CCAvg Education.1 Education.2 Education.3
## 1 40 10 84 2 2 0 1 0
## Mortgage SA CDA Online CreditCard
## 1 0 0 0 1 1
```

```

To_Predict_norm <- predict(Norm_model,To_Predict)
print(To_Predict_norm)

##           Age Experience      Income      Family      CCAvg Education.1 Educati
on.2
## 1 -0.4657003 -0.8811162 0.2221371 -0.3453975 0.0355115 -0.8494814      1.60
1024
## Education.3 Mortgage      SA      CDA      Online CreditCard
## 1 -0.6548999 -0.5554684 -0.3413892 -0.2535149 0.8218687      1.549477

Prediction <- knn(train = Train.df[1:13],
                  test = To_Predict_norm[1:13],
                  cl=Train.df$PL,
                  k=1)
print(Prediction)

## [1] 0
## Levels: 0 1

```

Addicting to the result, the customer will not be the targeted one.

Q2 What is a choice of k that balances between overfitting and ignoring the predictor information?

```

set.seed(123)
fitControl <- trainControl(method = "repeatedcv",
                           number = 3,
                           repeats = 2)
searchGrid=expand.grid(k = 1:15)
Knn.model=train(PL~.,
                data=Train.df,
                method='knn',
                tuneGrid=searchGrid,
                trControl = fitControl,)

Knn.model

## k-Nearest Neighbors
##
## 3000 samples
## 13 predictor
## 2 classes: '0', '1'
##
## No pre-processing
## Resampling: Cross-Validated (3 fold, repeated 2 times)
## Summary of sample sizes: 2000, 2000, 2000, 2000, 2000, 2000, ...
## Resampling results across tuning parameters:
##
##  k  Accuracy  Kappa
##  1  0.9561667  0.7225052
##  2  0.9531667  0.7043490

```

```
##      3  0.9573333  0.7078895
##      4  0.9553333  0.6929361
##      5  0.9558333  0.6916585
##      6  0.9548333  0.6848947
##      7  0.9521667  0.6592196
##      8  0.9496667  0.6340200
##      9  0.9488333  0.6274694
##     10  0.9480000  0.6163411
##     11  0.9480000  0.6139964
##     12  0.9473333  0.6082642
##     13  0.9470000  0.6048033
##     14  0.9438333  0.5748143
##     15  0.9441667  0.5757008
##
## Accuracy was used to select the optimal model using the largest value.
## The final value used for the model was k = 3.
```

Q3 Show the confusion matrix for the validation data that results from using the best k.

```
predictions<-predict(Knn.model,Validation.df)
confusionMatrix(predictions,Validation.df$PL)

## Confusion Matrix and Statistics
##
##              Reference
## Prediction    0    1
##              0 1800   73
##              1    8  119
##
##              Accuracy : 0.9595
##              95% CI : (0.9499, 0.9677)
##              No Information Rate : 0.904
##              P-Value [Acc > NIR] : < 2.2e-16
##
##              Kappa : 0.7251
##
##  Mcnemar's Test P-Value : 1.151e-12
##
##              Sensitivity : 0.9956
##              Specificity : 0.6198
##              Pos Pred Value : 0.9610
##              Neg Pred Value : 0.9370
##              Prevalence : 0.9040
##              Detection Rate : 0.9000
##              Detection Prevalence : 0.9365
##              Balanced Accuracy : 0.8077
##
##              'Positive' Class : 0
##
```

Q4 Consider the following customer: Age = 40, Experience = 10, Income = 84, Family = 2, CCAvg = 2, Education_1 = 0, Education_2 = 1, Education_3 = 0, Mortgage = 0, Securities Account = 0, CD Account = 0, Online = 1 and Credit Card = 1. Classify the customer using the best k.

```
To_Predict=data.frame(Age=40, Experience = 10, Income = 84, Family = 2, CCAvg
= 2,
                        Education.1 = 0, Education.2 = 1, Education.3 = 0, Mort
gage = 0,
                        SA = 0, CDA = 0, Online = 1,CreditCard = 1)
To_Predict_norm=predict(Norm_model,To_Predict)
predict(Knn.model,To_Predict_norm)

## [1] 0
## Levels: 0 1
```

Also, this customer is not the targeted one.