

# Analysis

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2022-04-26

## Import libraries and data

```
library(rpart)
German_credit <- read.csv("../Data_DA/GermanCredit.csv", header = TRUE, sep = ";")

# German_credit$DURATION <- as.numeric(German_credit$DURATION)
# German_credit$AMOUNT <- as.numeric(German_credit$AMOUNT)
# German_credit$INSTALL_RATE <- as.numeric(German_credit$INSTALL_RATE)
# German_credit$AGE <- as.numeric(German_credit$AGE)
# German_credit$NUM_CREDITS <- as.numeric(German_credit$NUM_CREDITS)
# German_credit$NUM_DEPENDENTS <- as.numeric(German_credit$NUM_DEPENDENTS)
#
# for (i in 1:ncol(German_credit)){
#   if (class(German_credit[,i])=="integer"){
#     German_credit[,i] <- factor(German_credit[,i])
#   }
# }
```

## Fitting a model :

Let's try a lassification tree

```
german.ct <- rpart(RESPONSE ~ ., method = "class", data = German_credit)
summary(german.ct)
```

```
## Call:
## rpart(formula = RESPONSE ~ ., data = German_credit, method = "class")
##   n= 1000
##
##          CP nsplit rel error   xerror   xstd
## 1 0.05166667    0 1.0000000 1.0000000 0.04830459
## 2 0.04666667    3 0.8400000 1.0066667 0.04839605
## 3 0.01833333    4 0.7933333 0.8800000 0.04646432
## 4 0.01400000    6 0.7566667 0.8600000 0.04612013
## 5 0.01333333   11 0.6866667 0.8633333 0.04617828
## 6 0.01000000   12 0.6733333 0.8966667 0.04674268
##
## Variable importance
##      CHK_ACCT      DURATION      AMOUNT      HISTORY      SAV_ACCT
##          30          14          10          10          9
##    REAL_ESTATE    USED_CAR      OBS.      AGE      RADIO_TV
##          5          4          4          3          3
##      JOB  PROP_UNKN_NONE  GUARANTOR  MALE_MAR_or_WID  EMPLOYMENT
```

```

##          2          1          1          1          1
##  INSTALL_RATE
##          1
##
## Node number 1: 1000 observations,    complexity param=0.05166667
##  predicted class=1  expected loss=0.3  P(node) =1
##    class counts:    300    700
##    probabilities: 0.300 0.700
##  left son=2 (543 obs) right son=3 (457 obs)
##  Primary splits:
##    CHK_ACCT < 1.5      to the left,  improve=47.90962, (0 missing)
##    HISTORY  < 1.5      to the left,  improve=17.06212, (0 missing)
##    SAV_ACCT < 1.5      to the left,  improve=14.80642, (0 missing)
##    DURATION < 34.5     to the right, improve=13.62155, (0 missing)
##    AMOUNT   < 3913.5   to the right, improve=11.32017, (0 missing)
##  Surrogate splits:
##    SAV_ACCT < 1.5      to the left,  agree=0.611, adj=0.149, (0 split)
##    HISTORY  < 3.5      to the left,  agree=0.592, adj=0.107, (0 split)
##    RADIO.TV < 0.5      to the left,  agree=0.565, adj=0.048, (0 split)
##    EMPLOYMENT < 3.5     to the left,  agree=0.554, adj=0.024, (0 split)
##    AGE      < 30.5     to the left,  agree=0.554, adj=0.024, (0 split)
##
## Node number 2: 543 observations,    complexity param=0.05166667
##  predicted class=1  expected loss=0.441989  P(node) =0.543
##    class counts:    240    303
##    probabilities: 0.442 0.558
##  left son=4 (237 obs) right son=5 (306 obs)
##  Primary splits:
##    DURATION < 22.5     to the right, improve=12.810640, (0 missing)
##    HISTORY  < 1.5      to the left,  improve= 9.653787, (0 missing)
##    REAL_ESTATE < 0.5   to the left,  improve= 9.181363, (0 missing)
##    SAV_ACCT < 1.5      to the left,  improve= 8.890786, (0 missing)
##    AMOUNT   < 8079     to the right, improve= 6.601270, (0 missing)
##  Surrogate splits:
##    AMOUNT   < 2805.5   to the right, agree=0.748, adj=0.422, (0 split)
##    PROP_UNKN_NONE < 0.5 to the right, agree=0.643, adj=0.181, (0 split)
##    USED_CAR  < 0.5     to the right, agree=0.599, adj=0.080, (0 split)
##    REAL_ESTATE < 0.5   to the left,  agree=0.597, adj=0.076, (0 split)
##    JOB       < 2.5     to the right, agree=0.595, adj=0.072, (0 split)
##
## Node number 3: 457 observations
##  predicted class=1  expected loss=0.131291  P(node) =0.457
##    class counts:    60    397
##    probabilities: 0.131 0.869
##
## Node number 4: 237 observations,    complexity param=0.05166667
##  predicted class=0  expected loss=0.4345992  P(node) =0.237
##    class counts:    134    103
##    probabilities: 0.565 0.435
##  left son=8 (196 obs) right son=9 (41 obs)
##  Primary splits:
##    SAV_ACCT < 2.5      to the left,  improve=7.374515, (0 missing)
##    USED_CAR  < 0.5      to the left,  improve=4.129437, (0 missing)
##    AMOUNT   < 1381.5   to the left,  improve=3.289316, (0 missing)

```

```

##      INSTALL_RATE < 2.5      to the right, improve=3.067516, (0 missing)
##      DURATION      < 43.5    to the right, improve=2.564920, (0 missing)
##
## Node number 5: 306 observations,      complexity param=0.04666667
## predicted class=1 expected loss=0.3464052 P(node) =0.306
## class counts:    106    200
## probabilities: 0.346 0.654
## left son=10 (28 obs) right son=11 (278 obs)
## Primary splits:
## HISTORY      < 1.5      to the left, improve=10.040510, (0 missing)
## OBS.         < 120.5    to the right, improve= 6.207418, (0 missing)
## REAL_ESTATE  < 0.5      to the left, improve= 5.585685, (0 missing)
## GUARANTOR    < 0.5      to the left, improve= 3.782059, (0 missing)
## DURATION     < 11.5     to the right, improve= 3.766531, (0 missing)
##
## Node number 8: 196 observations,      complexity param=0.01833333
## predicted class=0 expected loss=0.377551 P(node) =0.196
## class counts:    122    74
## probabilities: 0.622 0.378
## left son=16 (36 obs) right son=17 (160 obs)
## Primary splits:
## DURATION      < 47.5    to the right, improve=5.023838, (0 missing)
## USED_CAR      < 0.5      to the left, improve=4.598639, (0 missing)
## INSTALL_RATE  < 2.5      to the right, improve=2.682485, (0 missing)
## AMOUNT        < 11788    to the right, improve=2.516732, (0 missing)
## PRESENT_RESIDENT < 1.5    to the right, improve=1.984382, (0 missing)
## Surrogate splits:
## AMOUNT < 13319.5 to the right, agree=0.837, adj=0.111, (0 split)
##
## Node number 9: 41 observations
## predicted class=1 expected loss=0.2926829 P(node) =0.041
## class counts:    12    29
## probabilities: 0.293 0.707
##
## Node number 10: 28 observations
## predicted class=0 expected loss=0.25 P(node) =0.028
## class counts:    21    7
## probabilities: 0.750 0.250
##
## Node number 11: 278 observations,      complexity param=0.014
## predicted class=1 expected loss=0.3057554 P(node) =0.278
## class counts:    85    193
## probabilities: 0.306 0.694
## left son=22 (241 obs) right son=23 (37 obs)
## Primary splits:
## OBS.         < 120.5    to the right, improve=5.407923, (0 missing)
## AMOUNT        < 7491.5  to the right, improve=4.366338, (0 missing)
## DURATION     < 11.5     to the right, improve=3.840775, (0 missing)
## REAL_ESTATE  < 0.5      to the left, improve=3.589042, (0 missing)
## HISTORY      < 2.5      to the left, improve=2.954088, (0 missing)
##
## Node number 16: 36 observations
## predicted class=0 expected loss=0.1388889 P(node) =0.036
## class counts:    31    5

```

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##      probabilities: 0.861 0.139
##
## Node number 17: 160 observations,      complexity param=0.01833333
##      predicted class=0 expected loss=0.43125 P(node) =0.16
##      class counts:      91      69
##      probabilities: 0.569 0.431
##      left son=34 (137 obs) right son=35 (23 obs)
##      Primary splits:
##          USED_CAR      < 0.5      to the left, improve=5.092387, (0 missing)
##          AMOUNT        < 2313     to the left, improve=3.402464, (0 missing)
##          INSTALL_RATE  < 2.5      to the right, improve=2.374236, (0 missing)
##          NEW_CAR       < 0.5      to the right, improve=2.000321, (0 missing)
##          AGE           < 57.5     to the left, improve=1.711184, (0 missing)
##      Surrogate splits:
##          OBS. < 982.5   to the left, agree=0.863, adj=0.043, (0 split)
##          AGE < 62       to the left, agree=0.863, adj=0.043, (0 split)
##
## Node number 22: 241 observations,      complexity param=0.014
##      predicted class=1 expected loss=0.3443983 P(node) =0.241
##      class counts:      83      158
##      probabilities: 0.344 0.656
##      left son=44 (7 obs) right son=45 (234 obs)
##      Primary splits:
##          AMOUNT        < 7491.5   to the right, improve=3.790803, (0 missing)
##          OBS.          < 933.5     to the left, improve=3.525911, (0 missing)
##          GUARANTOR     < 0.5       to the left, improve=3.309626, (0 missing)
##          DURATION      < 11.5      to the right, improve=3.180698, (0 missing)
##          REAL_ESTATE   < 0.5       to the left, improve=2.854868, (0 missing)
##
## Node number 23: 37 observations
##      predicted class=1 expected loss=0.05405405 P(node) =0.037
##      class counts:      2      35
##      probabilities: 0.054 0.946
##
## Node number 34: 137 observations
##      predicted class=0 expected loss=0.379562 P(node) =0.137
##      class counts:      85      52
##      probabilities: 0.620 0.380
##
## Node number 35: 23 observations
##      predicted class=1 expected loss=0.2608696 P(node) =0.023
##      class counts:      6      17
##      probabilities: 0.261 0.739
##
## Node number 44: 7 observations
##      predicted class=0 expected loss=0.1428571 P(node) =0.007
##      class counts:      6      1
##      probabilities: 0.857 0.143
##
## Node number 45: 234 observations,      complexity param=0.014
##      predicted class=1 expected loss=0.3290598 P(node) =0.234
##      class counts:      77      157
##      probabilities: 0.329 0.671
##      left son=90 (200 obs) right son=91 (34 obs)

```

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## Primary splits:
##   DURATION < 8.5      to the right, improve=3.555963, (0 missing)
##   OBS.      < 933.5   to the left,  improve=3.224786, (0 missing)
##   GUARANTOR < 0.5     to the left,  improve=2.899666, (0 missing)
##   AMOUNT    < 1541.5  to the left,  improve=2.886843, (0 missing)
##   EDUCATION < 0.5     to the right, improve=2.874786, (0 missing)
## Surrogate splits:
##   AMOUNT < 527.5     to the right, agree=0.876, adj=0.147, (0 split)
##
## Node number 90: 200 observations,      complexity param=0.014
##   predicted class=1 expected loss=0.365 P(node) =0.2
##   class counts:      73   127
##   probabilities: 0.365 0.635
##   left son=180 (85 obs) right son=181 (115 obs)
## Primary splits:
##   AMOUNT          < 1423   to the left,  improve=4.928926, (0 missing)
##   OBS.            < 933.5  to the left,  improve=3.101534, (0 missing)
##   GUARANTOR       < 0.5    to the left,  improve=2.315746, (0 missing)
##   PRESENT_RESIDENT < 3.5    to the left,  improve=2.108112, (0 missing)
##   MALE_SINGLE     < 0.5    to the left,  improve=1.895859, (0 missing)
## Surrogate splits:
##   INSTALL_RATE    < 3.5    to the right, agree=0.655, adj=0.188, (0 split)
##   DURATION        < 12.5   to the left,  agree=0.640, adj=0.153, (0 split)
##   JOB             < 1.5    to the left,  agree=0.625, adj=0.118, (0 split)
##   MALE_MAR_or_WID < 0.5    to the right, agree=0.605, adj=0.071, (0 split)
##   AGE            < 45     to the right, agree=0.605, adj=0.071, (0 split)
##
## Node number 91: 34 observations
##   predicted class=1 expected loss=0.1176471 P(node) =0.034
##   class counts:      4    30
##   probabilities: 0.118 0.882
##
## Node number 180: 85 observations,      complexity param=0.014
##   predicted class=1 expected loss=0.4941176 P(node) =0.085
##   class counts:      42   43
##   probabilities: 0.494 0.506
##   left son=360 (48 obs) right son=361 (37 obs)
## Primary splits:
##   REAL_ESTATE     < 0.5    to the left,  improve=6.566190, (0 missing)
##   NUM_CREDITS     < 1.5    to the left,  improve=5.195552, (0 missing)
##   GUARANTOR       < 0.5    to the left,  improve=4.715579, (0 missing)
##   PRESENT_RESIDENT < 3.5    to the left,  improve=4.440830, (0 missing)
##   AGE            < 37.5   to the left,  improve=2.973182, (0 missing)
## Surrogate splits:
##   RADIO_TV        < 0.5    to the left,  agree=0.729, adj=0.378, (0 split)
##   GUARANTOR       < 0.5    to the left,  agree=0.659, adj=0.216, (0 split)
##   JOB            < 1.5    to the right, agree=0.659, adj=0.216, (0 split)
##   AMOUNT         < 632    to the right, agree=0.624, adj=0.135, (0 split)
##   MALE_MAR_or_WID < 0.5    to the left,  agree=0.624, adj=0.135, (0 split)
##
## Node number 181: 115 observations
##   predicted class=1 expected loss=0.2695652 P(node) =0.115
##   class counts:      31   84
##   probabilities: 0.270 0.730

```

```

##
## Node number 360: 48 observations,    complexity param=0.01333333
##   predicted class=0   expected loss=0.3333333   P(node) =0.048
##   class counts:      32    16
##   probabilities: 0.667 0.333
##   left son=720 (34 obs) right son=721 (14 obs)
##   Primary splits:
##     AGE                < 37.5    to the left,   improve=3.787115, (0 missing)
##     NUM_CREDITS         < 1.5     to the left,   improve=3.555556, (0 missing)
##     PRESENT_RESIDENT    < 2.5     to the left,   improve=2.711485, (0 missing)
##     AMOUNT              < 967     to the left,   improve=1.864802, (0 missing)
##     EMPLOYMENT          < 1.5     to the left,   improve=1.434174, (0 missing)
##   Surrogate splits:
##     EDUCATION           < 0.5     to the left,   agree=0.750, adj=0.143, (0 split)
##     NUM_CREDITS         < 1.5     to the left,   agree=0.750, adj=0.143, (0 split)
##     JOB                 < 0.5     to the right,  agree=0.750, adj=0.143, (0 split)
##     NUM_DEPENDENTS      < 1.5     to the left,   agree=0.750, adj=0.143, (0 split)
##     PRESENT_RESIDENT    < 3.5     to the left,   agree=0.729, adj=0.071, (0 split)
##
## Node number 361: 37 observations
##   predicted class=1   expected loss=0.2702703   P(node) =0.037
##   class counts:      10    27
##   probabilities: 0.270 0.730
##
## Node number 720: 34 observations
##   predicted class=0   expected loss=0.2058824   P(node) =0.034
##   class counts:      27     7
##   probabilities: 0.794 0.206
##
## Node number 721: 14 observations
##   predicted class=1   expected loss=0.3571429   P(node) =0.014
##   class counts:       5     9
##   probabilities: 0.357 0.643

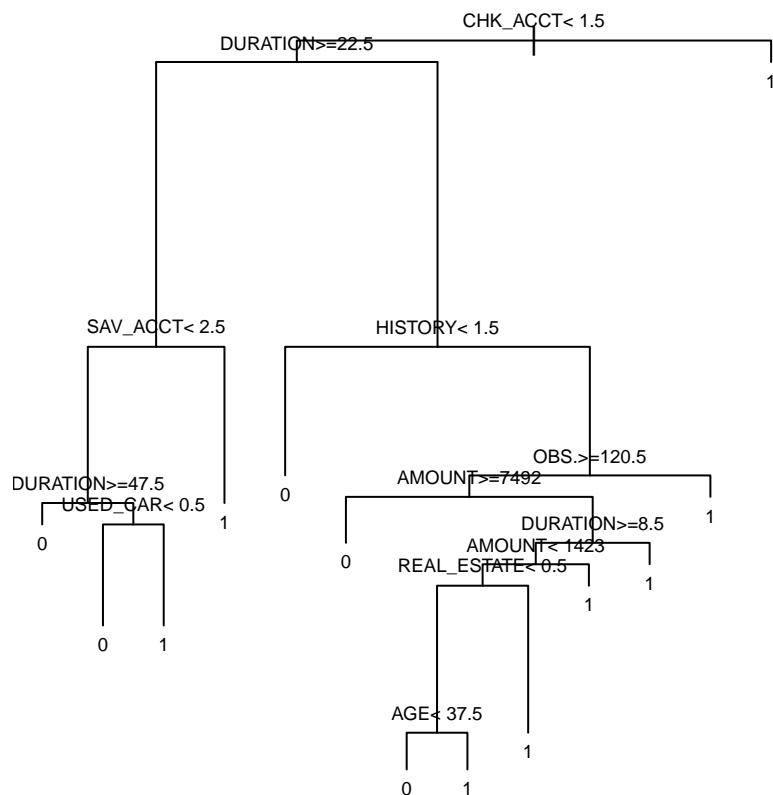
```

The model is not working yet.

```

par(pty = "s", mar = c(1, 1, 1, 1))
plot(german.ct, cex = 1)
text(german.ct, cex = 0.6)

```



Fitting another model :

```

# Logistic regression to see the significant variables (not working)
mod1 <- glm(RESPONSE~., data = German_credit, family= binomial)
summary(mod1)

##
## Call:
## glm(formula = RESPONSE ~ ., family = binomial, data = German_credit)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.6301  -0.7228   0.3889   0.7030   2.3722
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  1.057e+00  8.704e-01   1.214  0.22479
## OBS.        -1.623e-04  2.884e-04  -0.563  0.57356
## CHK_ACCT     5.633e-01  7.254e-02   7.766 8.12e-15 ***
## DURATION    -2.719e-02  9.046e-03  -3.005  0.00265 **
## HISTORY      3.968e-01  8.980e-02   4.419 9.93e-06 ***
## NEW_CAR     -7.996e-01  3.811e-01  -2.098  0.03588 *
## USED_CAR      8.218e-01  4.782e-01   1.718  0.08571 .
## FURNITURE    -4.223e-02  3.952e-01  -0.107  0.91490
## RADIO.TV      6.264e-02  3.846e-01   0.163  0.87061
## EDUCATION    -9.249e-01  4.952e-01  -1.868  0.06182 .
## RETRAINING   -8.732e-02  4.376e-01  -0.200  0.84182
## AMOUNT      -1.168e-04  4.272e-05  -2.734  0.00625 **

```

```

## SAV_ACCT          2.497e-01  6.066e-02  4.116 3.85e-05 ***
## EMPLOYMENT        1.168e-01  7.469e-02  1.564 0.11781
## INSTALL_RATE      -3.171e-01  8.659e-02 -3.662 0.00025 ***
## MALE_DIV          -3.443e-01  3.814e-01 -0.903 0.36663
## MALE_SINGLE        5.378e-01  2.051e-01  2.622 0.00874 **
## MALE_MAR_or_WID    1.069e-01  3.046e-01  0.351 0.72572
## CO.APPLICANT       -3.494e-01  3.989e-01 -0.876 0.38113
## GUARANTOR          9.451e-01  4.146e-01  2.279 0.02265 *
## PRESENT_RESIDENT  -1.242e-02  8.403e-02 -0.148 0.88247
## REAL_ESTATE        1.997e-01  2.096e-01  0.953 0.34083
## PROP_UNKN_NONE     -5.569e-01  3.735e-01 -1.491 0.13595
## AGE                1.211e-02  8.383e-03  1.445 0.14843
## OTHER_INSTALL      -6.310e-01  2.045e-01 -3.085 0.00203 **
## RENT               -6.277e-01  4.608e-01 -1.362 0.17313
## OWN_RES            -2.222e-01  4.360e-01 -0.510 0.61030
## NUM_CREDITS        -2.238e-01  1.663e-01 -1.346 0.17833
## JOB               -3.325e-02  1.427e-01 -0.233 0.81569
## NUM_DEPENDENTS     -2.480e-01  2.461e-01 -1.008 0.31349
## TELEPHONE          3.507e-01  1.955e-01  1.794 0.07288 .
## FOREIGN            1.458e+00  6.243e-01  2.336 0.01951 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##    Null deviance: 1221.73  on 999  degrees of freedom
## Residual deviance:  907.75  on 968  degrees of freedom
## AIC: 971.75
##
## Number of Fisher Scoring iterations: 5

```