

# Risk Classification

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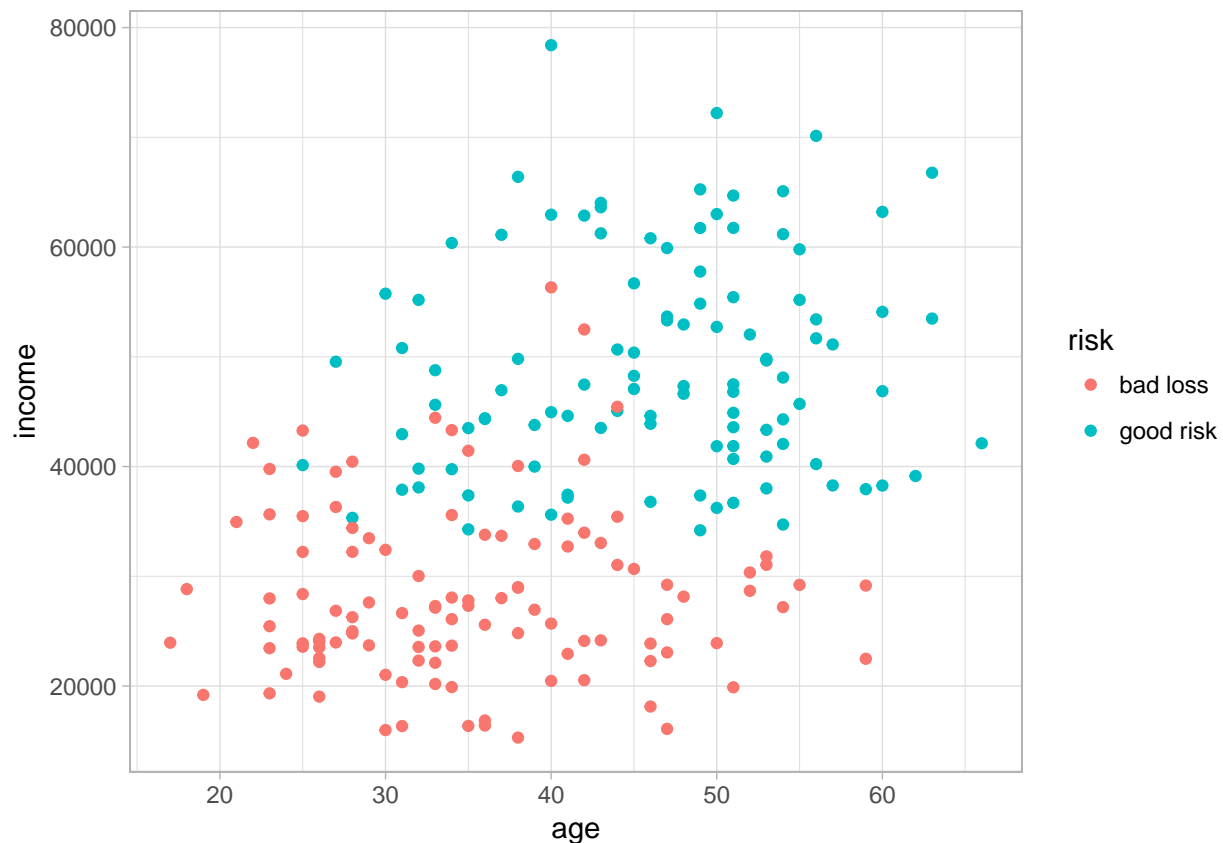
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```
rm(list=ls())
setwd("~/IS665/risk")

classify_risk = read.csv("ClassifyRisk_historical.csv", header=TRUE, sep=",", dec=".")

require(ggplot2)

ggplot(data=classify_risk) + geom_point(aes(x=age, y=income, color=risk)) + theme_light()
```



```
require(class)
require(magrittr)

summary(classify_risk)
```

##	mortgage	loans	age	marital_status	income
##	n: 64	Min. :0.000	Min. :17.00	married:65	Min. :15301
##	y:153	1st Qu.:1.000	1st Qu.:32.00	other :53	1st Qu.:26857
##		Median :1.000	Median :40.00	single :99	Median :37363
##		Mean :1.336	Mean :40.54		Mean :38401
##		3rd Qu.:2.000	3rd Qu.:50.00		3rd Qu.:47335

```

##           Max.      :3.000   Max.      :66.00           Max.      :78399
##           risk
##    bad loss :111
##    good risk:106
##
##
##
##

n.classify_risk <- data.frame(sapply(classify_risk[,c(2,3,5)], function(x) {
(x - min(x))/(max(x) - min(x))
}))

head(n.classify_risk)

##           loans      age      income
## 1 0.33333333 0.3877551 0.2929044
## 2 0.33333333 0.7346939 0.4444643
## 3 0.33333333 0.9387755 0.6051203
## 4 0.33333333 0.6122449 0.6025608
## 5 0.66666667 0.6734694 0.3317385
## 6 0.00000000 0.6938776 0.7360926

# Split
set.seed(1234) # makes it repeatable
ind <- sample(2, nrow(classify_risk), replace = TRUE, prob = c(0.67, 0.33))
## Data Split
classify_risk.training <- n.classify_risk[ind == 1, 1:3]
classify_risk.test <- n.classify_risk[ind == 2, 1:3]

n.classify_risk[, "risk"] <- classify_risk[, "risk"]

## Label Split
classify_risk.trainLabels <- n.classify_risk[ind == 1, 4]
classify_risk.testLabels <- n.classify_risk[ind == 2, 4]

classify_risk_pred <- knn(train = classify_risk.training, test =
                           classify_risk.test, cl =
                           classify_risk.trainLabels, k = 5)

results = data.frame(classify_risk_pred, classify_risk.testLabels)
table(results)

##           classify_risk.testLabels
## classify_risk_pred bad loss good risk
##           bad loss      27      1
##           good risk      2     32

# install.packages('caret')
require(caret)
# install.packages('e1071')
require(e1071)
confusionMatrix(table(results))

## Confusion Matrix and Statistics
##
##           classify_risk.testLabels

```

```

## classify_risk_pred bad loss good risk
##      bad loss      27      1
##      good risk      2     32
##
##      Accuracy : 0.9516
##      95% CI : (0.865, 0.9899)
##      No Information Rate : 0.5323
##      P-Value [Acc > NIR] : 2.844e-13
##
##      Kappa : 0.9026
##      McNemar's Test P-Value : 1
##
##      Sensitivity : 0.9310
##      Specificity : 0.9697
##      Pos Pred Value : 0.9643
##      Neg Pred Value : 0.9412
##      Prevalence : 0.4677
##      Detection Rate : 0.4355
##      Detection Prevalence : 0.4516
##      Balanced Accuracy : 0.9504
##
##      'Positive' Class : bad loss
##
test_classify_risk <- read.csv("classifyrisk.csv", header=TRUE, sep=",", dec=".")

n.test_classify_risk <- data.frame(sapply(test_classify_risk[,c(2,3,5)], function(x) {
  (x - min(x))/(max(x) - min(x))
}))

classify_risk_pred.test <- n.test_classify_risk
head(classify_risk_pred.test)

##      loans      age      income
## 1 0.3333333 0.8139535 0.6121850
## 2 0.0000000 0.5813953 0.6242710
## 3 0.3333333 0.4418605 0.8957524
## 4 0.3333333 0.8372093 0.7006266
## 5 0.0000000 0.3720930 0.6875295
## 6 0.0000000 0.5581395 0.8846504

classify_test_result <- knn(train = classify_risk.training, test = classify_risk_pred.test,
  cl = classify_risk[ind == 1, 6], k = 5)

(classify_test_result)

## [1] good risk good risk good risk good risk good risk good risk bad loss
## [8] bad loss bad loss good risk bad loss good risk good risk good risk
## [15] good risk bad loss good risk good risk bad loss bad loss good risk
## [22] good risk good risk good risk bad loss bad loss bad loss good risk
## [29] bad loss
## Levels: bad loss good risk

test_classify_risk[, "risk"] <- classify_test_result

test_classify_risk

```

##	mortgage	loans	age	marital_status	income	risk
## 1	n	1	54	married	50203.25	good risk
## 2	y	0	44	married	50793.46	good risk
## 3	y	1	38	single	64051.12	good risk
## 4	y	1	55	married	54522.25	good risk
## 5	y	0	35	single	53882.66	good risk
## 6	y	0	43	single	63508.96	good risk
## 7	n	2	62	married	27030.50	bad loss
## 8	y	3	41	other	20307.50	bad loss
## 9	y	1	26	single	24777.08	bad loss
## 10	y	0	50	married	48931.66	good risk
## 11	n	2	30	other	23575.22	bad loss
## 12	y	0	54	married	53242.06	good risk
## 13	n	1	48	single	39527.75	good risk
## 14	y	0	37	married	51852.40	good risk
## 15	y	1	43	single	69142.00	good risk
## 16	y	2	23	other	22039.40	bad loss
## 17	y	1	34	married	54180.75	good risk
## 18	y	1	54	married	53011.50	good risk
## 19	y	2	33	single	35558.00	bad loss
## 20	n	1	19	married	20954.50	bad loss
## 21	y	0	49	married	50873.56	good risk
## 22	y	2	48	single	39826.60	good risk
## 23	y	3	55	other	39417.88	good risk
## 24	y	1	38	married	52561.75	good risk
## 25	y	2	24	single	32626.00	bad loss
## 26	y	1	23	single	24314.50	bad loss
## 27	y	1	39	single	23785.64	bad loss
## 28	n	1	52	married	35380.75	good risk
## 29	n	1	49	single	29352.50	bad loss