

project

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```
library(tidyr)
library(ggplot2)
library(randomForest)
```

```
## randomForest 4.7-1.1
```

```
## Type rfNews() to see new features/changes/bug fixes.
```

```
##
## Attaching package: 'randomForest'
```

```
## The following object is masked from 'package:ggplot2':
##
##     margin
```

```
library(caret)
```

```
## Loading required package: lattice
```

```
library(glmnet)
```

```
## Loading required package: Matrix
```

```
##
## Attaching package: 'Matrix'
```

```
## The following objects are masked from 'package:tidyr':
##
##     expand, pack, unpack
```

```
## Loaded glmnet 4.1-6
```

```
## Data exploratory
bank = read.csv('data/bank-additional/bank-additional-full.csv', sep = ';')
summary(bank)
```

```

##      age      job      marital      education
## Min.   :17.00 Length:41188 Length:41188 Length:41188
## 1st Qu.:32.00 Class :character Class :character Class :character
## Median :38.00 Mode  :character Mode  :character Mode  :character
## Mean   :40.02
## 3rd Qu.:47.00
## Max.   :98.00
##      default      housing      loan      contact
## Length:41188 Length:41188 Length:41188 Length:41188
## Class :character Class :character Class :character Class :character
## Mode  :character Mode  :character Mode  :character Mode  :character
##
##
##      month      day_of_week      duration      campaign
## Length:41188 Length:41188 Min.   : 0.0 Min.   : 1.000
## Class :character Class :character 1st Qu.: 102.0 1st Qu.: 1.000
## Mode  :character Mode  :character Median : 180.0 Median : 2.000
## Mean   : 258.3 Mean   : 2.568
## 3rd Qu.: 319.0 3rd Qu.: 3.000
## Max.   :4918.0 Max.   :56.000
##      pdays      previous      poutcome      emp.var.rate
## Min.   : 0.0 Min.   :0.000 Length:41188 Min.   : -3.40000
## 1st Qu.:999.0 1st Qu.:0.000 Class :character 1st Qu.: -1.80000
## Median :999.0 Median :0.000 Mode  :character Median : 1.10000
## Mean   :962.5 Mean   :0.173 Mean   : 0.08189
## 3rd Qu.:999.0 3rd Qu.:0.000 3rd Qu.: 1.40000
## Max.   :999.0 Max.   :7.000 Max.   : 1.40000
## cons.price.idx cons.conf.idx euribor3m nr.employed
## Min.   :92.20 Min.   : -50.8 Min.   :0.634 Min.   :4964
## 1st Qu.:93.08 1st Qu.: -42.7 1st Qu.:1.344 1st Qu.:5099
## Median :93.75 Median : -41.8 Median :4.857 Median :5191
## Mean   :93.58 Mean   : -40.5 Mean   :3.621 Mean   :5167
## 3rd Qu.:93.99 3rd Qu.: -36.4 3rd Qu.:4.961 3rd Qu.:5228
## Max.   :94.77 Max.   : -26.9 Max.   :5.045 Max.   :5228
##      y
## Length:41188
## Class :character
## Mode  :character
##
##
##

```

```
names(bank)
```

```

## [1] "age"      "job"      "marital"  "education"
## [5] "default"  "housing"  "loan"     "contact"
## [9] "month"    "day_of_week" "duration" "campaign"
## [13] "pdays"   "previous"  "poutcome" "emp.var.rate"
## [17] "cons.price.idx" "cons.conf.idx" "euribor3m" "nr.employed"
## [21] "y"

```

```

# remove predictors
# default a extremely unbalanced
# duration is unknown while predicting
remove.variable = c('default', 'duration')
bank = bank[,!names(bank) %in% remove.variable]

dummy.variables = c('job', 'marital', 'education',
                    'housing', 'loan', 'contact',
                    'month', 'day_of_week', 'poutcome',
                    'y')

for(name in dummy.variables) {
  bank[name] = as.factor(bank[[name]])
}
summary(bank)

```

```

##      age      job      marital
##  Min.   :17.00  admin.   :10422  divorced: 4612
##  1st Qu.:32.00  blue-collar: 9254  married :24928
##  Median :38.00  technician : 6743  single  :11568
##  Mean   :40.02  services   : 3969  unknown :   80
##  3rd Qu.:47.00  management : 2924
##  Max.   :98.00  retired    : 1720
##              (Other)   : 6156
##
##      education      housing      loan      contact
##  university.degree :12168  no      :18622  no      :33950  cellular :26144
##  high.school        : 9515  unknown:  990  unknown:  990  telephone:15044
##  basic.9y           : 6045  yes      :21576  yes      : 6248
##  professional.course: 5243
##  basic.4y           : 4176
##  basic.6y           : 2292
##  (Other)            : 1749
##
##      month      day_of_week      campaign      pdays      previous
##  may      :13769  fri:7827  Min.   : 1.000  Min.   : 0.0  Min.   :0.000
##  jul      : 7174  mon:8514  1st Qu.: 1.000  1st Qu.:999.0  1st Qu.:0.000
##  aug      : 6178  thu:8623  Median : 2.000  Median :999.0  Median :0.000
##  jun      : 5318  tue:8090  Mean    : 2.568  Mean    :962.5  Mean    :0.173
##  nov      : 4101  wed:8134  3rd Qu.: 3.000  3rd Qu.:999.0  3rd Qu.:0.000
##  apr      : 2632              Max.    :56.000  Max.    :999.0  Max.    :7.000
##  (Other): 2016
##
##      poutcome      emp.var.rate      cons.price.idx      cons.conf.idx
##  failure   : 4252  Min.   :-3.40000  Min.   :92.20  Min.   : -50.8
##  nonexistent:35563  1st Qu.: -1.80000  1st Qu.:93.08  1st Qu.: -42.7
##  success    : 1373  Median : 1.10000  Median :93.75  Median : -41.8
##              Mean    : 0.08189  Mean    :93.58  Mean    : -40.5
##              3rd Qu.: 1.40000  3rd Qu.:93.99  3rd Qu.: -36.4
##              Max.    : 1.40000  Max.    :94.77  Max.    : -26.9
##
##      euribor3m      nr.employed      y
##  Min.   :0.634  Min.   :4964  no :36548
##  1st Qu.:1.344  1st Qu.:5099  yes: 4640
##  Median :4.857  Median :5191

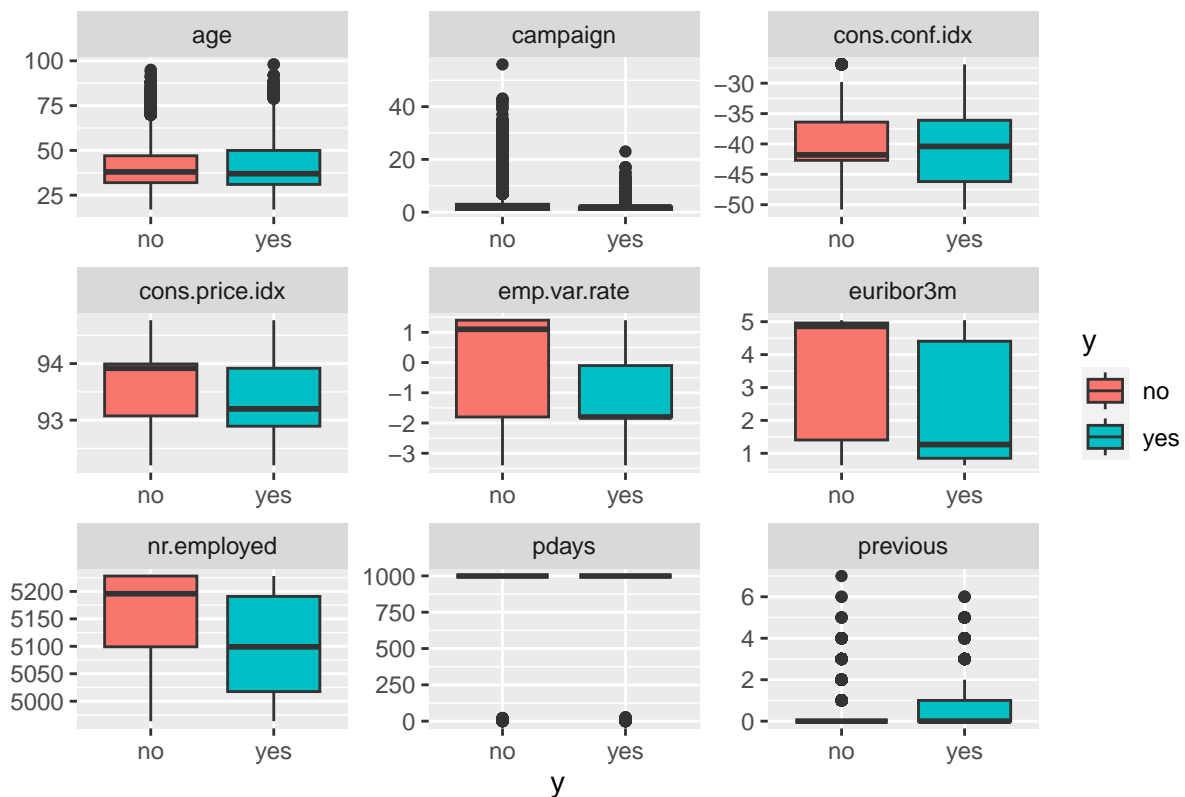
```

```
## Mean :3.621 Mean :5167
## 3rd Qu.:4.961 3rd Qu.:5228
## Max. :5.045 Max. :5228
##
```

```
#bank.scaled = bank
# scale all continuous predictors
#for (name in setdiff(names(bank), dummy.variables)) {
# bank.scaled[name] = scale(bank.scaled[name])
#}
```

```
bank[!(names(bank) %in% dummy.variables[1:9])] %>%
  gather(-y, key = 'var', value = 'value') %>%
  ggplot(aes(x = y, y = value)) +
  geom_boxplot(aes(fill = y)) +
  facet_wrap(~var, scales = 'free') +
  ylab('')+
  ggtitle('Predictors Distribution against Diabetes')
```

Predictors Distribution against Diabetes



```
# add column: if the client was last contacted from a previous campaign
#bank$ispcontacted = as.factor((bank$pdays == 999))

# missing value only account for 8% of the whole data
# no additional process on missing value, treat them as a category
miss.name = 'unknown'
print('missing values')
```

```
## [1] "missing values"
```

```
for (name in dummy.variables) {  
  print(paste(name, sum(bank[name] == miss.name), sep=':'))  
}
```

```
## [1] "job:330"  
## [1] "marital:80"  
## [1] "education:1731"  
## [1] "housing:990"  
## [1] "loan:990"  
## [1] "contact:0"  
## [1] "month:0"  
## [1] "day_of_week:0"  
## [1] "poutcome:0"  
## [1] "y:0"
```

```
# total records with missing values  
sum(rowSums(bank==miss.name)>0)
```

```
## [1] 2943
```

```
pr.curve = function(pred, y) {  
  recalls = c()  
  precisions = c()  
  
  for (threshold in seq(0, 1, by=0.01)) {  
    yhat = ifelse(pred>threshold, 'yes', 'no')  
    yhat = factor(yhat, levels = c('no', 'yes'), labels = c('no', 'yes'))  
    recalls = c(recalls, recall(yhat, y, relevant = 'yes'))  
    precisions = c(precisions, precision(yhat, y, relevant = 'yes'))  
  }  
  data.frame(recall = recalls, precision = precisions, threshold=seq(0, 1, by=0.01))  
}
```

```
set.seed(5420)
```

```
# separate to train and test  
train.size = 2*nrow(bank)/3  
train.rows = sample(nrow(bank), train.size)
```

```
rf = randomForest(y~., data = bank, subset = train.rows, importance = TRUE)  
rf
```

```
##  
## Call:  
## randomForest(formula = y ~ ., data = bank, importance = TRUE, subset = train.rows)  
##           Type of random forest: classification  
##           Number of trees: 500  
## No. of variables tried at each split: 4  
##
```

```
##          OOB estimate of  error rate: 10.43%
## Confusion matrix:
##          no yes class.error
## no  23669 642    0.0264078
## yes  2223 924    0.7063870
```

```
yhat.rf = predict(rf, newdata = bank[-train.rows,], type='prob')
#confusionMatrix(yhat.norm, bank[-train.rows, 'y'])
pr.rf = pr.curve(yhat.rf[, 'yes'], bank[-train.rows, 'y'])

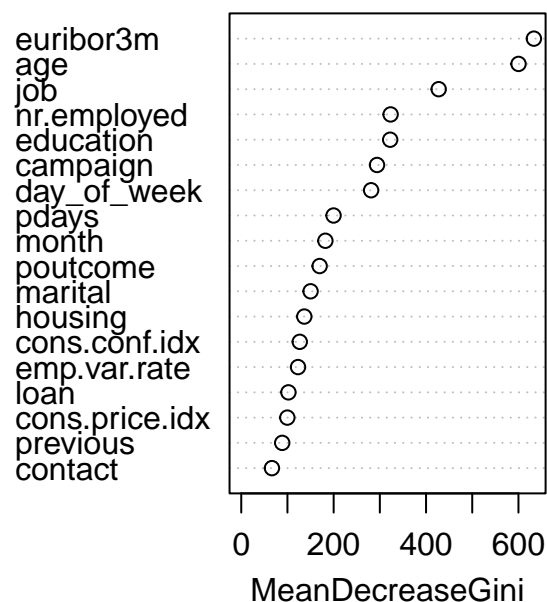
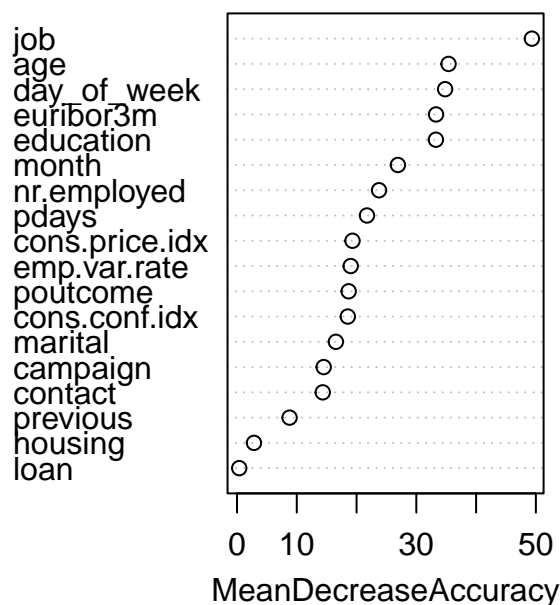
pr.rf$type = 'RF'
```

```
importance(rf)
```

	no	yes	MeanDecreaseAccuracy	MeanDecreaseGini
## age	36.2958035	-3.79177710	35.3859048	600.14474
## job	56.7230287	-14.01171493	49.3340220	427.35221
## marital	17.9019050	-2.58654343	16.5430319	150.10480
## education	34.1963001	0.50296773	33.2733214	322.35301
## housing	3.3929888	-0.34948738	2.8375034	136.48883
## loan	0.3730673	0.06219239	0.3740652	102.14547
## contact	11.1040111	25.52766861	14.3441883	66.43997
## month	26.2315182	-10.94829139	26.9175663	182.11714
## day_of_week	36.3727999	-1.07743292	34.8188780	281.08233
## campaign	10.2859199	10.42018414	14.4988762	293.88781
## pdays	8.1561100	26.50234707	21.7410649	199.78561
## previous	7.8032107	3.24009428	8.7924984	88.68205
## poutcome	13.6088217	14.04864784	18.6786793	169.68921
## emp.var.rate	18.3148838	2.22817635	19.0233140	122.89921
## cons.price.idx	19.2723364	-13.49077101	19.3202683	99.75588
## cons.conf.idx	18.2262713	-9.06997109	18.5368083	126.67364
## euribor3m	31.0114341	6.41700313	33.2980190	633.41846
## nr.employed	21.0422642	15.78838362	23.7608870	323.32416

```
varImpPlot(rf)
```

rf



```
## LR stepwise
trControl = trainControl(method = 'cv', number = 10)
lr.step = train(y~.,
  data = bank[train.rows,],
  method = 'glmStepAIC',
  family = 'binomial',
  direction = 'backward',
  trControl = trControl,
  trace = F
)
summary(lr.step$finalModel)
```

```
##
## Call:
## NULL
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.1173  -0.3901  -0.3257  -0.2674   3.0167
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)   -3.110e+02  3.234e+01  -9.617  < 2e-16 ***
## 'jobblue-collar' -1.387e-01  6.624e-02  -2.094  0.036227 *
## jobmanagement  -1.512e-01  8.884e-02  -1.702  0.088703 .
## jobretired      2.263e-01  8.937e-02   2.533  0.011325 *
```

```

## jobstudent          2.387e-01  1.127e-01   2.118 0.034183 *
## maritalsingle       8.790e-02  4.936e-02   1.781 0.074969 .
## educationbasic.9y  -1.600e-01  7.457e-02  -2.145 0.031920 *
## educationuniversity.degree 1.155e-01  5.072e-02  2.276 0.022825 *
## contacttelephone    -6.945e-01  7.767e-02  -8.942 < 2e-16 ***
## monthaug           6.793e-01  1.136e-01   5.981 2.21e-09 ***
## monthdec           7.301e-01  2.227e-01   3.279 0.001042 **
## monthjun          -8.244e-01  1.171e-01  -7.040 1.92e-12 ***
## monthmar           1.809e+00  1.435e-01  12.602 < 2e-16 ***
## monthmay          -3.212e-01  7.515e-02  -4.274 1.92e-05 ***
## monthnov          -2.450e-01  8.560e-02  -2.862 0.004205 **
## monthoct           3.360e-01  1.263e-01   2.661 0.007780 **
## monthsep           5.824e-01  1.652e-01   3.525 0.000423 ***
## day_of_weekmon     -2.069e-01  5.646e-02  -3.665 0.000247 ***
## day_of_weekwed      1.071e-01  5.524e-02   1.939 0.052492 .
## campaign          -4.640e-02  1.120e-02  -4.142 3.45e-05 ***
## pdays             -1.029e-03  2.292e-04  -4.491 7.10e-06 ***
## poutcomenonexistent  4.990e-01  6.941e-02   7.188 6.56e-13 ***
## poutcomesuccess     8.003e-01  2.302e-01   3.477 0.000508 ***
## emp.var.rate       -1.688e+00  1.466e-01 -11.509 < 2e-16 ***
## cons.price.idx      2.571e+00  2.326e-01  11.053 < 2e-16 ***
## cons.conf.idx       3.574e-02  5.480e-03   6.521 7.00e-11 ***
## nr.employed         1.365e-02  2.116e-03   6.451 1.11e-10 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
## Null deviance: 19553 on 27457 degrees of freedom
## Residual deviance: 15330 on 27431 degrees of freedom
## AIC: 15384
##
## Number of Fisher Scoring iterations: 6

```

```

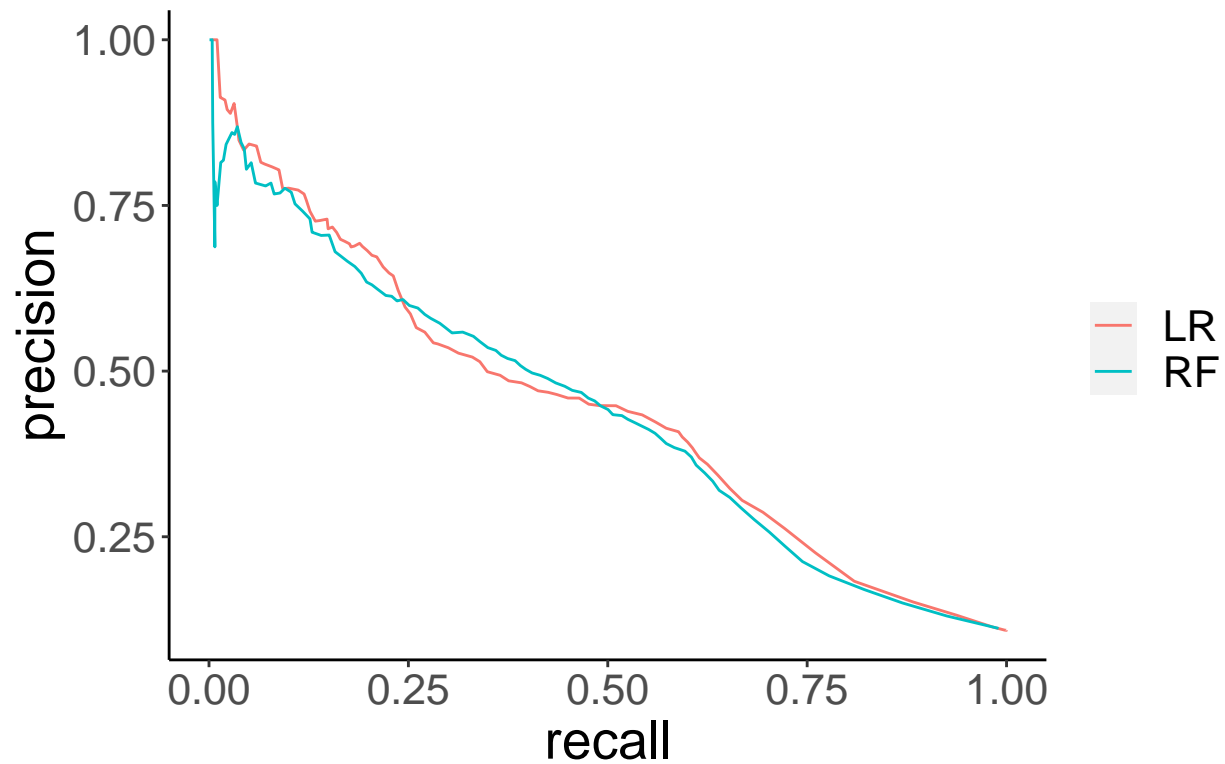
yhat.lr.step = predict.train(lr.step,
                             newdata = bank[-train.rows,],
                             type="prob")
pr.lr.step = pr.curve(yhat.lr.step[, 'yes'], bank[-train.rows, 'y'])
pr.lr.step$type = 'LR'

# plot pr curve
data = rbind(pr.rf, pr.lr.step)
ggplot(data = data)+
  geom_line(aes(x = recall, y=precision, col = type))+
  ggtitle('Precision-Recall Curve')+
  labs(color = NULL)+
  theme(text = element_text(size = 20),
        panel.background = element_blank(),
        axis.line = element_line(colour = "black")
  )

```

```
## Warning: Removed 14 rows containing missing values ('geom_line()').
```


Precision–Recall Curve



```
# F1 score
data$f1 = 2*data$recall*data$precision/(data$recall+data$precision)
ggplot(data = data)+
  geom_line(aes(x = recall, y=f1, col=type))+
  ggtitle('F1 Score')+
  labs(color = NULL)+
  theme(text = element_text(size = 20),
        panel.background = element_blank(),
        axis.line = element_line(colour = "black")
  )
```

```
## Warning: Removed 14 rows containing missing values ('geom_line()').
```

