Model Fitting II

Carolina Alves de Lima Salge

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Slide Code

```
library(tidyverse)
## -- Attaching packages ----- tidyverse 1.3.0 --
## v ggplot2 3.3.3
                             0.3.4
                   v purrr
## v tibble 3.1.0
                  v dplyr
                            1.0.4
                   v stringr 1.4.0
## v tidyr
          1.1.2
## v readr
          1.4.0
                   v forcats 0.5.1
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                  masks stats::lag()
library(caret)
## Loading required package: lattice
## Attaching package: 'caret'
## The following object is masked from 'package:purrr':
##
##
      lift
insurance <- read_csv("insurance.csv")</pre>
##
## -- Column specification -----
## cols(
    age = col_double(),
    sex = col_character(),
##
##
    bmi = col_double(),
    children = col_double(),
##
    smoker = col_character(),
    region = col_character(),
    charges = col_double()
##
## )
```

```
## # A tibble: 1,338 x 7
##
      age sex
                  bmi children smoker region
                                          charges
##
     <dbl> <chr> <dbl> <chr> <dbl> <chr> <
                                              <dbl>
## 1
     19 female 27.9
                           0 yes
                                   southwest 16885.
## 2 18 male 33.8
                           1 no
                                   southeast
                                            1726.
## 3 28 male 33
                          3 no
                                   southeast 4449.
## 4 33 male
               22.7
                          0 no
                                   northwest 21984.
                          0 no
## 5 32 male
                 28.9
                                   northwest 3867.
## 6 31 female 25.7
                          0 no
                                   southeast 3757.
## 7 46 female 33.4
                          1 no
                                   southeast 8241.
## 8
       37 female 27.7
                                   northwest 7282.
                           3 no
## 9
       37 male
                 29.8
                           2 no
                                   northeast 6406.
## 10
       60 female 25.8
                           0 no
                                   northwest 28923.
## # ... with 1,328 more rows
# transform categories to numbers
insurance <- insurance %>%
 mutate(sexN = case_when(
   sex == "male" ~ 1,
   sex == "female" ~ 0
   )) %>%
 mutate(smokerN = case_when(
   smoker == "yes" ~ 1,
   smoker == "no" ~ 0
   )) %>%
 mutate(regionN = case_when(
   region == "southwest" ~ 1,
   region == "southeast" ~ 2,
   region == "northwest" ~ 3,
   region == "northwest" ~ 4
   ))
# only select numeric variables
df <- insurance %>%
 dplyr::select(charges, age, sexN, bmi, children, smokerN, regionN)
# drop missing values NAs
df1 <- drop_na(df)</pre>
# compute correlation between predictos
cor(df1[,2:7])
##
                                              children
                  age
                            sexN
                                        bmi
                                                         smokerN
## age
           -0.019119318 1.000000000 0.06887488 0.011989296 0.07964590
## sexN
           ## children 0.058405316 0.011989296 0.01926722 1.000000000 0.01040992
## smokerN -0.021043239 0.079645898 0.02191144 0.010409918 1.00000000
## regionN -0.007367464 -0.004927365 -0.09209306 0.002040754 0.00000000
              regionN
         -0.007367464
## age
```

```
## sexN
           -0.004927365
## bmi
           -0.092093057
## children 0.002040754
## smokerN 0.00000000
## regionN 1.00000000
# compute correlation between predictos and the target
cor(df1[,1:7])
##
               charges
                                                         bmi
                                                                children
                                age
                                            sexN
## charges 1.000000000 0.298743694 0.062371225 0.19260362 0.068287662
           0.298743694 \quad 1.000000000 \quad -0.019119318 \quad 0.09570218 \quad 0.058405316
## age
           0.062371225 -0.019119318 1.000000000 0.06887488 0.011989296
## sexN
           ## bmi
## children 0.068287662 0.058405316 0.011989296 0.01926722 1.000000000
## smokerN 0.802212827 -0.021043239 0.079645898 0.02191144 0.010409918
## regionN 0.002286109 -0.007367464 -0.004927365 -0.09209306 0.002040754
##
               smokerN
                            regionN
## charges 0.80221283 0.002286109
## age
           -0.02104324 -0.007367464
## sexN
            0.07964590 -0.004927365
## bmi
            0.02191144 -0.092093057
## children 0.01040992 0.002040754
## smokerN 1.00000000 0.000000000
## regionN 0.0000000 1.00000000
# age, bmi, and smoking are highly correlated with health costs
set.seed(12L) # set a starting seed to be able to get reproducible results
# partition data
trainIndex <- createDataPartition(df1$charges, # target variable</pre>
                                 p = 0.8, # percentage that goes to training
                                 list = FALSE, # results will not be in a list
                                 times = 1) # number of partitions to create
charges_train <- df1[trainIndex, ] # data frame for training</pre>
charges_test <- df1[-trainIndex, ] # data frame for testing</pre>
# use training set to build model
model <- train(charges ~ age + bmi + smokerN,</pre>
              data = charges_train, # use training set
              method = "lm") # linear regression
# now predict outcomes in test set
p <- predict(model, charges_test)</pre>
# how did we do? calculate performance across resamples
# RMSE and R-squared
postResample(pred = p, obs = charges_test$charges)
##
          RMSE
                  Rsquared
                  0.7916937 4016.8698553
```

5779.1881082

```
# on average, our prediction is off by $5,779.18
# how can we improve performance? Try a different method!
model2 <- train(charges ~ age + bmi + smokerN,</pre>
               data = charges_train, # use training set
               method = "ranger") # random forest
## note: only 2 unique complexity parameters in default grid. Truncating the grid to 2 .
# now predict outcomes in test set
p1 <- predict(model2, charges_test)</pre>
# how did we do? calculate performance across resamples
# RMSE and R-squared
postResample(pred = p1, obs = charges_test$charges)
##
           RMSE
                    Rsquared
                                      MAF.
## 4603.8843625
                   0.8683862 2625.9694526
# on average, our prediction is off by $4,632.99
# first collect the resampling results of each model
resamps <- resamples(list(LM = model,
                          RF = model2)
resamps
##
## Call:
## resamples.default(x = list(LM = model, RF = model2))
## Models: LM, RF
## Number of resamples: 25
## Performance metrics: MAE, RMSE, Rsquared
## Time estimates for: everything, final model fit
# then use a simple t-test to evaluate the null hypothesis that there is no difference
summary(diff(resamps))
##
## Call:
## summary.diff.resamples(object = diff(resamps))
##
## p-value adjustment: bonferroni
## Upper diagonal: estimates of the difference
## Lower diagonal: p-value for HO: difference = 0
##
## MAE
##
     LM
                RF
## LM
                1491
## RF < 2.2e-16
##
```

RMSE

LM RF ## LM 1220

RF 4.433e-13

Rsquared

LM RF ## LM -0.08187

RF 5.742e-11