Model Fitting II

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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 3.0.6 v dplyr
                            1.0.4
## v tidyr
          1.1.2
                   v stringr 1.4.0
## 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.
        60 female 25.8
## 10
                             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 == "northeast" ~ 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
           1.000000000 -0.020855872 0.109271882 0.04246900 -0.025018752
          -0.020855872 \quad 1.000000000 \quad 0.046371151 \quad 0.01716298 \quad 0.076184817
## sexN
           0.003750426
## children 0.042468999 0.017162978 0.012758901 1.00000000 0.007673120
## smokerN -0.025018752 0.076184817 0.003750426 0.00767312 1.000000000
## regionN -0.002127313 -0.004588385 -0.157565849 -0.01656945 0.002180682
               regionN
          -0.002127313
## age
```

```
## sexN
           -0.004588385
## bmi
           -0.157565849
## children -0.016569446
## smokerN 0.002180682
## regionN 1.00000000
# compute correlation between predictos and the target
cor(df1[,1:7])
##
                                                         bmi
                                                                children
               charges
                                age
                                           sexN
## charges 1.000000000 0.299008193 0.057292062 0.198340969 0.06799823
           0.299008193 1.000000000 -0.020855872 0.109271882 0.04246900
## age
           0.057292062 -0.020855872 1.000000000 0.046371151
## sexN
                                                              0.01716298
## bmi
           ## children 0.067998227 0.042468999 0.017162978 0.012758901 1.00000000
## smokerN 0.787251430 -0.025018752 0.076184817 0.003750426 0.00767312
## regionN 0.006208235 -0.002127313 -0.004588385 -0.157565849 -0.01656945
##
                smokerN
                             regionN
## charges 0.787251430 0.006208235
## age
           -0.025018752 -0.002127313
## sexN
            0.076184817 -0.004588385
## bmi
            0.003750426 -0.157565849
## children 0.007673120 -0.016569446
## smokerN 1.00000000 0.002180682
## regionN 0.002180682 1.000000000
# 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>
## Warning: The 'i' argument of ''['()' can't be a matrix as of tibble 3.0.0.
## Convert to a vector.
## This warning is displayed once every 8 hours.
## Call 'lifecycle::last_warnings()' to see where this warning was generated.
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)
##
                    Rsquared
           RMSE
                                      MAE
## 5808.0045894
                   0.7989742 4184.9721150
# on average, our prediction is off by $5,808.00
# how can we improve performance? Try a different method!
model2 <- train(charges ~ age + bmi + smokerN,</pre>
               data = charges_train, # use training set
               method = "ranger") # random forest
## note: only 2 unique complexity parameters in default grid. Truncating the grid to 2 .
# now predict outcomes in test set
p1 <- predict(model2, charges_test)</pre>
# how did we do? calculate performance across resamples
# RMSE and R-squared
postResample(pred = p1, obs = charges_test$charges)
           RMSE
                    Rsquared
## 4205.4282707
                   0.8933816 2436.1432025
# 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
                1403
## RF < 2.2e-16
##
## RMSE
##
                RF
     LM
## LM
                1156
## RF 2.139e-15
##
## Rsquared
##
     LM
               RF
## LM
               -0.09466
## RF 1.17e-14
```