UWDS2-Wk6-LinearModeling-Abalone-jms206

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Due 24 Feb 2015

Overview

This assignment uses the abalone dataset from the UCI machine learning data repository. Each row in the dataset describes a different abalone, including its sex, linear dimensions, and weights. The last column contains the number of rings in an abalone's shell. This is a proxy for the abalone's age, just as tree rings tell us how old a tree is. The problem we face is coming up with an easy to apply model that predicts the number of rings (counting them is laborious and pretty unpleasant for the abalone, which gets sawed in half).

The data is available here: (http://archive.ics.uci.edu/ml/machine-learning-databases/abalone/abalone.data)

The data description is here: (http://archive.ics.uci.edu/ml/machine-learning-databases/abalone/abalone.names)

Put the data.frame into a data.table in order to simplify column references below.

```
library(data.table)
abodt <- data.table(abo)
summary(abodt)</pre>
```

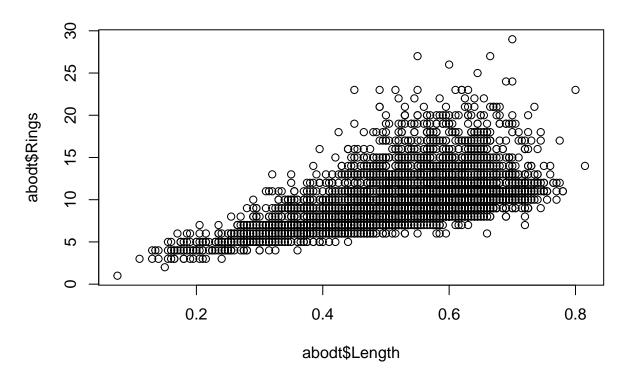
```
##
   Sex
                 Length
                               Diameter
                                                 Height
##
   F:1307
            Min.
                   :0.075
                            Min.
                                   :0.0550
                                             Min.
                                                    :0.0000
   I:1342
            1st Qu.:0.450
                            1st Qu.:0.3500
                                             1st Qu.:0.1150
##
##
   M:1528
            Median :0.545
                            Median :0.4250
                                             Median :0.1400
                                    :0.4079
                                                    :0.1395
##
            Mean
                   :0.524
                            Mean
                                             Mean
            3rd Qu.:0.615
                            3rd Qu.:0.4800
                                             3rd Qu.:0.1650
##
            Max.
##
                  :0.815
                            Max.
                                   :0.6500
                                             Max.
                                                    :1.1300
##
    Whole_weight
                    Shucked_weight Viscera_weight
                                                       Shell_weight
          :0.0020
                            :0.0010 Min.
                                                             :0.0015
##
   Min.
                    Min.
                                            :0.0005
                                                      Min.
##
    1st Qu.:0.4415
                    1st Qu.:0.1860
                                     1st Qu.:0.0935
                                                      1st Qu.:0.1300
   Median :0.7995
                    Median :0.3360 Median :0.1710
                                                      Median :0.2340
##
   Mean
          :0.8287
                    Mean
                           :0.3594
                                     Mean
                                            :0.1806
                                                      Mean
                                                             :0.2388
##
##
   3rd Ou.:1.1530
                    3rd Ou.:0.5020
                                     3rd Ou.:0.2530
                                                      3rd Ou.:0.3290
##
   Max.
          :2.8255
                    Max.
                           :1.4880
                                     Max.
                                            :0.7600
                                                      Max.
                                                             :1.0050
##
       Rings
```

Min. : 1.000 ## 1st Qu.: 8.000 ## Median : 9.000 ## Mean : 9.934 ## 3rd Qu.:11.000 ## Max. :29.000

1. Plot Rings vs Length

Plot the number of rings as a function of length.

plot(abodt\$Length, abodt\$Rings)



2. Linear Model: Rings ~ Length

__ Fit a linear model to this data (rings = a*length + b) using R's Im command. Examine the output of the summary table for the fit. Is length a significant factor? __

```
abolm <- lm(Rings ~ Length, data=abodt)
abolm_summary <- summary(abolm)
abolm_summary</pre>
```

```
##
## Call:
## lm(formula = Rings ~ Length, data = abodt)
##
## Residuals:
```

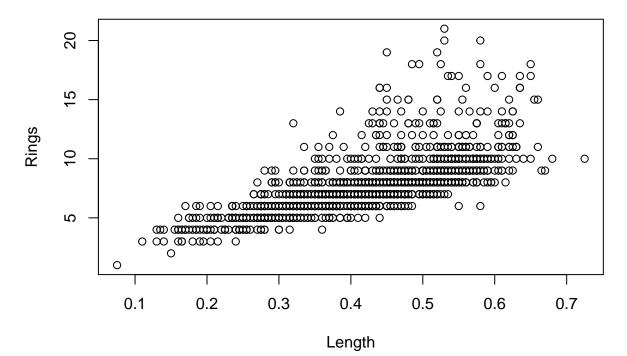
```
##
       Min
                1Q Median
                                3Q
                                       Max
##
  -5.9665 -1.6961 -0.7423 0.8733 16.6776
##
## Coefficients:
               Estimate Std. Error t value Pr(>|t|)
##
                 2.1019
                            0.1855
                                     11.33
                                              <2e-16 ***
## (Intercept)
                14.9464
                                     43.30
                                              <2e-16 ***
## Length
                            0.3452
                   0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Signif. codes:
##
## Residual standard error: 2.679 on 4175 degrees of freedom
## Multiple R-squared: 0.3099, Adjusted R-squared:
## F-statistic: 1875 on 1 and 4175 DF, p-value: < 2.2e-16
```

With a p-value $< 2^15$, length is a significant factor.

3. Linear Model, Immature Abalone: Rings ~ Length

__There are three sexes of abalone: male, female, and immature. Filter the data so that only the immature abalone remain. Fit the same model to this data (rings = a*length + b). Examine the output of summary: is this model a better or worse than the model fit to all of the data?___

```
immaturedt <- abodt[abodt$Sex == "I",]
with (immaturedt, plot(Length, Rings))</pre>
```



```
immaturelm1 <- lm(Rings ~ Length, data=immaturedt)
immaturelm1_summary <- summary(immaturelm1)
immaturelm1_summary</pre>
```

```
##
## Call:
## lm(formula = Rings ~ Length, data = immaturedt)
##
## Residuals:
##
      Min
               10 Median
                                3Q
                                       Max
## -4.3002 -1.0736 -0.3398 0.5271 11.4911
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
                            0.2024
                                     5.535 3.73e-08 ***
## (Intercept) 1.1204
                            0.4586 34.513 < 2e-16 ***
## Length
               15.8273
## ---
## Signif. codes:
                   0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.828 on 1340 degrees of freedom
## Multiple R-squared: 0.4706, Adjusted R-squared: 0.4702
## F-statistic: 1191 on 1 and 1340 DF, p-value: < 2.2e-16
stopifnot(abolm_summary$adj.r.squared < immaturelm1_summary$adj.r.squared)</pre>
```

"Goodness of fit" is usually defined by R^2 , or the fraction of variance in the data that's explained by the model.

The adjusted R^2 value is used here, rather than R^2 , in preparation for comparing this value to that obtained when the number of predictors is increased (e.g. height and diameter). Unadjusted R^2 can be inflated by the mere presence of additional predictors. Adjusted R^2 takes the number of predictors into account.

The adjusted \mathbb{R}^2 value for linear model of immature abalone of 0.47 is notably larger than the corresponding value of 0.31 for the linear model of the entire abolone sample (female, male, and immature).

In summary, a linear model using Length as a sole predictor of rings is a better predictor when applied to the immature (pre-adult) subset of abolone than when applied to the entire dataset.

4. Linear Model, Immature Abalone: Rings ~ Length + Height + Diameter

Still working with the immature abalone only, add Height and Diameter to the model (rings = alength + bheight + etc.). Examine the output of summary: what are the significant factors in this new model? Compare the result to the "length only" model and explain why the two results on consistent with each other.

```
immaturelm2 <- lm(Rings ~ Length + Height + Diameter, data=immaturedt)
immaturelm2_summary <- summary(immaturelm2)
immaturelm2_summary

##
## Call:
## lm(formula = Rings ~ Length + Height + Diameter, data = immaturedt)</pre>
```

##

```
## Residuals:
##
      Min
               1Q Median
                                3Q
                                      Max
  -4.4320 -1.0075 -0.2701 0.5536 11.0270
##
##
## Coefficients:
##
              Estimate Std. Error t value Pr(>|t|)
                1.5223
                         0.1976
                                    7.703 2.57e-14 ***
## (Intercept)
## Length
               -2.9797
                           2.6292 -1.133 0.25728
                           3.6467 11.190 < 2e-16 ***
## Height
               40.8082
## Diameter
                9.9103
                           3.3316
                                    2.975 0.00299 **
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.729 on 1338 degrees of freedom
## Multiple R-squared: 0.5274, Adjusted R-squared: 0.5263
## F-statistic: 497.7 on 3 and 1338 DF, p-value: < 2.2e-16
# Assert expected p-values for the three predictors:
stopifnot(immaturelm2_summary$coefficients["Length","Pr(>|t|)"] > 0.25)
stopifnot(immaturelm2_summary$coefficients["Height","Pr(>|t|)"] < 0.05)</pre>
stopifnot(immaturelm2_summary$coefficients["Diameter","Pr(>|t|)"] < 0.05)</pre>
```

The significant predictors (significant: a p-value less than 0.05) are *Height* and *Diameter* (with p-values of 0.000 and 0.003, respectively. *Length* is not significant (p-value of 0.26 is greater than 0.05).

Compare the result to the 'length only' model and explain why the two results on consistent with each other.

The fit with three predictors (two, effectively, because of Length's high p-value) is better than with one predictor: 0.53 vs 0.47.

But why has *Length* been displaced? By itself, it's a pretty good predictor, yet when *Diameter* and *Length* are added, it becomes insignificant (high p-value)?

Reason: Because Length is strongly correlated with Diameter (0.986).

Estimate Std. Error t value Pr(>|t|)

##

##

Coefficients:

Sanity Check: Use Length with Diameter should yield a similar good fit as using Height with Diameter:

```
immaturelm3 <- lm(Rings ~ Length + Height, data=immaturedt)
immaturelm3_summary <- summary(immaturelm3)
immaturelm3_summary

##
## Call:
## lm(formula = Rings ~ Length + Height, data = immaturedt)
##
## Residuals:
## Min    1Q Median    3Q    Max
## -4.5666 -1.0215 -0.2717    0.5584    11.0690</pre>
```

```
## (Intercept)
                1.3914
                            0.1932
                                   7.201 9.92e-13 ***
## Length
                4.2061
                            1.0408 4.041 5.62e-05 ***
                            3.5413 12.289 < 2e-16 ***
## Height
               43.5199
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.734 on 1339 degrees of freedom
## Multiple R-squared: 0.5243, Adjusted R-squared: 0.5235
## F-statistic: 737.8 on 2 and 1339 DF, p-value: < 2.2e-16
stopifnot(abs(immaturelm3_summary$adj.r.squared - immaturelm2_summary$adj.r.squared) < 0.01)</pre>
```

The $adjustedR^2$ values are within 0.01 of each other.

The following use of "." caused this error:

5. Linear Model, Immature Abalone: Rings ~ . -Sex

Still working with the immature abalone only, add all of the factors to the model (except Sex: since we only have immature abalone, this value is the same for every data point) (rings = alength + bheight + etc.).

```
# Error in `contrasts<-`(`*tmp*`, value = contr.funs[1 + isOF[nn]]) :</pre>
# contrasts can be applied only to factors with 2 or more levels
##mmaturelm4 <- lm(Rings ~ .-Sex, data=immaturedt)</pre>
# Explicitly specifying all the predictors works:
immaturelm4 <- lm(Rings ~ Length + Diameter + Height + Whole_weight + Shucked_weight + Viscera_weight
immaturelm4_summary <- summary(immaturelm4)</pre>
immaturelm4_summary
##
## Call:
## lm(formula = Rings ~ Length + Diameter + Height + Whole_weight +
##
       Shucked_weight + Viscera_weight + Shell_weight, data = immaturedt)
##
## Residuals:
       Min
                10 Median
##
                                3Q
                                       Max
## -8.0870 -0.9375 -0.2514 0.5947 10.0078
##
## Coefficients:
##
                  Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                    2.9380
                               0.3113 9.437 < 2e-16 ***
## Length
                               2.4851 -1.105 0.269347
                   -2.7461
## Diameter
                    5.9345
                               3.1786 1.867 0.062116 .
## Height
                   28.8253
                               3.6232 7.956 3.77e-15 ***
## Whole_weight
                               1.4910 5.518 4.12e-08 ***
                    8.2272
## Shucked_weight -14.6625
                               1.6283 -9.005 < 2e-16 ***
## Viscera_weight -11.2791
                               3.2068 -3.517 0.000451 ***
```

```
## Shell_weight
                   10.6606
                               2.3282
                                        4.579 5.11e-06 ***
##
                   0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.616 on 1334 degrees of freedom
## Multiple R-squared: 0.5879, Adjusted R-squared: 0.5858
## F-statistic: 271.9 on 7 and 1334 DF, p-value: < 2.2e-16
# Using all predictors yields a somewhat higher adjusted R^2:
stopifnot((immaturelm4_summary$adj.r.squared - immaturelm2_summary$adj.r.squared) > 0.05)
stopifnot(abs(immaturelm4_summary$adj.r.squared - immaturelm2_summary$adj.r.squared) < 0.1)</pre>
```

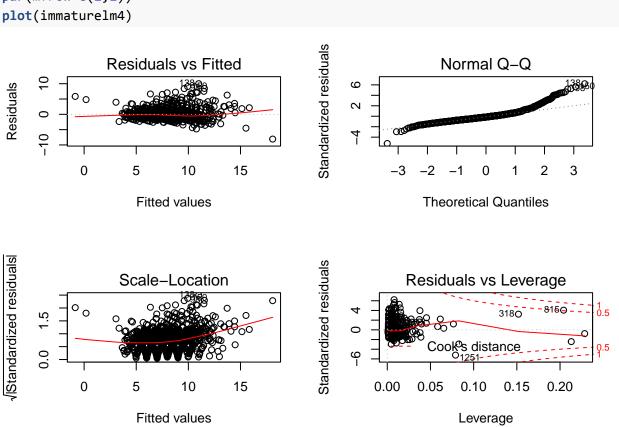
Examine the residuals and summarize your observations.

All predictors were significant except Length and Diameter.

The $adjustedR^2$ with all predictors of 0.586 is somewhat but not hugely better than the $adjustedR^2$ for Length+Height + Diameter of 0.526.

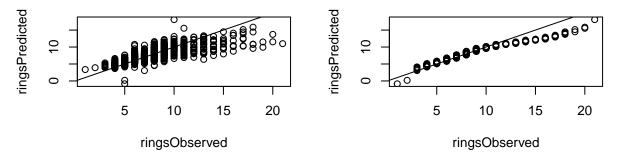
Use graphical methods (histogram, qqplot, etc.)

```
par(mfrow=c(2,2))
```



```
ringsPredicted <- predict(immaturelm4)</pre>
ringsObserved <- immaturedt$Rings</pre>
plot(ringsObserved, ringsPredicted, main="Scatter Plot: Rings Observed vs Predicted")
abline(0, 1)
qqplot(ringsObserved, ringsPredicted, main="Q-Q Plot: Rings Observed vs Predicted")
abline(0, 1)
```

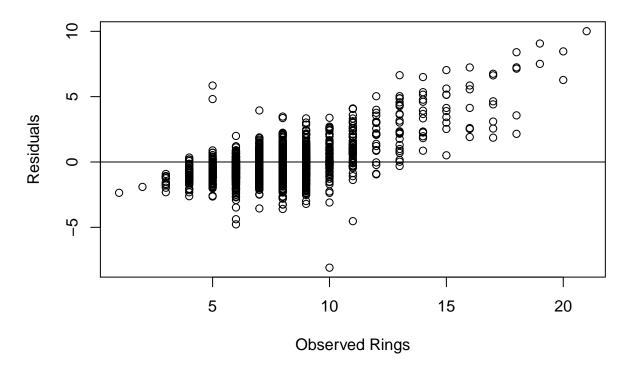
Scatter Plot: Rings Observed vs Predict Q-Q Plot: Rings Observed vs Predicte



Also plot the residuals as a function of the number of rings.

Assumption: 'number of rings' means 'observed (not predicted) number of rings'"

Residuals as Function of Observed Number of Rings



Residuals as Function of Predicted Number of Rings

