Tutorial 4: Regression Model Estimation

Anh Le (anh.le@duke.edu)
September 18, 2015

Agenda

- 1. Create data frames
- 2. Subset data frames
- 3. Estimate a linear model with lm()
- 4. Tips and tricks
- Prefix your objects in R (and related TAB tricks, i.e. arguments within function, variables within a data frame)
- fig.height(), fig.width()
- Code length <= 80 (RStudio > Tools > Options > Code)

1. Create data frames

```
## var1 var2 var3
## 1 11 21 a
## 2 12 22 b
## 3 13 23 c
```

2. Subset data frames

All subsetting can be done with the following construct: my_dataframe[?1 , ?2]

The first question mark (?1) refers to which rows we want. The second question mark (?2) refers to which columns we want.

How to indicate to R which rows / columns we want? Multiple ways:

1. Use rows / columns index

```
my_dataframe[1, 2]
```

```
## [1] 21
```

```
my_dataframe[1:2, 2]
## [1] 21 22
my_dataframe[1:2, ]
##
     var1 var2 var3
            21
## 1
       11
## 2
       12
            22
                   b
Rapid fire quiz
my_dataframe[2:3, ]
my_dataframe[ , 1:2]
my_dataframe[1:2, 2:3]
my_dataframe[c(1, 3), ]
my_dataframe[c(1, 3, 2), ]
  2. Use rows / columns name
my_dataframe[ , "var2"]
## [1] 21 22 23
Rapid fire quiz:
my_dataframe[ , c("var1", "var3")]
my_dataframe[c(2, 3), c("var1", "var3")]
  3. Use a condition
my_dataframe[c(TRUE, TRUE, FALSE), ]
     var1 var2 var3
## 1
            21
       11
       12
            22
my_dataframe[, c(TRUE, FALSE, TRUE)]
##
     var1 var3
## 1
       11
## 2
       12
             b
## 3
       13
             С
```

Of course this is not tenable for a large data frame. So we have this very useful trick:

```
my_dataframe[my_dataframe$var1 < 13, ]</pre>
##
     var1 var2 var3
## 1
            21
       11
       12
            22
                   b
This works because my_dataframe$var1 < 13 actuall returns c(TRUE, TRUE, FALSE) (vectorized operation
in the wild!). Indeed:
my_dataframe$var1 < 13
## [1] TRUE TRUE FALSE
Rapid fire quiz:
my_dataframe[my_dataframe$var2 == 22, ]
my_dataframe[my_dataframe$var2 == 25, ]
  4. Use a combination of condition
my_dataframe[my_dataframe$var1 > 10 & my_dataframe$var2 > 21, ]
##
     var1 var2 var3
       12
            22
                   b
            23
## 3
       13
my_dataframe[my_dataframe$var1 > 10 | my_dataframe$var2 > 21, ]
##
     var1 var2 var3
## 1
       11
            21
## 2
       12
            22
                   b
## 3
       13
            23
                   С
```

3. Estimate a linear model with lm()

In this section, I'll demo a (simplified) pipeline of steps in doing regression analysis with real data.

Download and clean data

```
library(WDI)
## Loading required package: RJSONIO
```

There are a lot of unwanted columns. What if I just want country, year, and the three variables of interest (NY.GDP.PCAP.CD, SP.DYN.IMRT.IN, SH.MED.PHYS.ZS)? (Hint: subsetting)

Rename columns:

```
colnames(d_2010)

## [1] "country" "year" "NY.GDP.PCAP.CD" "SP.DYN.IMRT.IN"

## [5] "SH.MED.PHYS.ZS"

colnames(d_2010)[3:5] <- c('gdppc', 'infant_mortality', 'number_of_physician')
colnames(d_2010)

## [1] "country" "year" "gdppc"

## [4] "infant_mortality" "number_of_physician"

Log gdp per capita

d_2010$log_gdppc <- log(d_2010$gdppc)</pre>
```

Build a linear model

```
lm(infant_mortality ~ log_gdppc, data = d_2010)
##
## Call:
## lm(formula = infant_mortality ~ log_gdppc, data = d_2010)
## Coefficients:
## (Intercept)
                  log_gdppc
##
        139.31
                     -13.14
m1 <- lm(infant_mortality ~ log_gdppc, data = d_2010)</pre>
summary(m1)
##
## Call:
## lm(formula = infant_mortality ~ log_gdppc, data = d_2010)
## Residuals:
```

```
1Q Median
                               3Q
                                      Max
                            6.222 79.971
## -27.641 -10.393 -2.596
##
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
                           6.8043
                                    20.47
## (Intercept) 139.3147
                           0.7909 -16.61
## log_gdppc
              -13.1403
                                           <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 16.14 on 184 degrees of freedom
    (31 observations deleted due to missingness)
## Multiple R-squared: 0.6, Adjusted R-squared: 0.5979
## F-statistic: 276 on 1 and 184 DF, p-value: < 2.2e-16
m2 <- lm(infant_mortality ~ log_gdppc + number_of_physician, data = d_2010)
summary(m2)
##
## lm(formula = infant_mortality ~ log_gdppc + number_of_physician,
##
      data = d_2010)
##
## Residuals:
##
      Min
               1Q Median
                               3Q
                                      Max
## -27.343 -8.182 -1.260
                            5.961 50.581
##
## Coefficients:
##
                      Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                      121.1074
                                   7.4312 16.297 < 2e-16 ***
                                   0.9824 -10.774 < 2e-16 ***
## log_gdppc
                      -10.5850
## number_of_physician -2.9102
                                   0.9288 -3.133 0.00211 **
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 12.71 on 140 degrees of freedom
     (74 observations deleted due to missingness)
## Multiple R-squared: 0.7078, Adjusted R-squared: 0.7036
## F-statistic: 169.6 on 2 and 140 DF, p-value: < 2.2e-16
```

Extract result from the model

\$ coefficients : Named num [1:2] 139.3 -13.1

str() (stands for structure) is used to look into the structure of an object in R, see what it contains.

```
str(m1)
## List of 13
```

```
## ..- attr(*, "names")= chr [1:2] "(Intercept)" "log_gdppc"
## $ residuals : Named num [1:186] 3.36 5.05 18.96 -7.12 -15.22 ...
## ..- attr(*, "names")= chr [1:186] "6" "7" "8" "9" ...
## $ effects : Named num [1:186] -381.68 268.13 19.23 -7.54 -15.39 ...
```

```
..- attr(*, "names")= chr [1:186] "(Intercept)" "log_gdppc" "" "" ...
##
                  : int 2
   $ rank
## $ fitted.values: Named num [1:186] -0.864 2.251 56.135 14.823 30.022 ...
    ..- attr(*, "names")= chr [1:186] "6" "7" "8" "9" ...
##
   $ assign
                  : int [1:2] 0 1
                  :List of 5
## $ gr
    ..$ qr : num [1:186, 1:2] -13.6382 0.0733 0.0733 0.0733 0.0733 ...
    ... - attr(*, "dimnames")=List of 2
##
    .. .. ..$ : chr [1:186] "6" "7" "8" "9" ...
##
    .....$ : chr [1:2] "(Intercept)" "log_gdppc"
     .. ..- attr(*, "assign")= int [1:2] 0 1
##
     ..$ qraux: num [1:2] 1.07 1.09
##
    ..$ pivot: int [1:2] 1 2
##
    ..$ tol : num 1e-07
##
    ..$ rank : int 2
##
    ..- attr(*, "class")= chr "qr"
   $ df.residual : int 184
##
                 :Class 'omit' Named int [1:31] 9 12 14 24 45 46 61 65 72 78 ...
    ....- attr(*, "names")= chr [1:31] "14" "17" "NA" "29" ...
##
## $ xlevels
                 : Named list()
## $ call
                  : language lm(formula = infant_mortality ~ log_gdppc, data = d_2010)
                 :Classes 'terms', 'formula' length 3 infant_mortality ~ log_gdppc
## $ terms
    ... - attr(*, "variables")= language list(infant_mortality, log_gdppc)
##
    .. ..- attr(*, "factors")= int [1:2, 1] 0 1
##
    .. .. - attr(*, "dimnames")=List of 2
##
    ..... s: chr [1:2] "infant_mortality" "log_gdppc"
##
    .. .. ... : chr "log_gdppc"
    .. ..- attr(*, "term.labels")= chr "log_gdppc"
##
    ...- attr(*, "order")= int 1
##
    .. ..- attr(*, "intercept")= int 1
    ....- attr(*, "response")= int 1
##
##
    ....- attr(*, ".Environment")=<environment: R_GlobalEnv>
    ... ..- attr(*, "predvars")= language list(infant_mortality, log_gdppc)
##
     ....- attr(*, "dataClasses")= Named chr [1:2] "numeric" "numeric"
##
    ..... attr(*, "names")= chr [1:2] "infant_mortality" "log_gdppc"
##
                  :'data.frame':
                                 186 obs. of 2 variables:
##
   $ model
    ..$ infant_mortality: num [1:186] 2.5 7.3 75.1 7.7 14.8 ...
##
                       : num [1:186] 10.67 10.43 6.33 9.47 8.32 ...
    ..$ log_gdppc
##
    ..- attr(*, "terms")=Classes 'terms', 'formula' length 3 infant_mortality ~ log_gdppc
##
    ..... attr(*, "variables")= language list(infant_mortality, log_gdppc)
    .. .. - attr(*, "factors")= int [1:2, 1] 0 1
##
     ..... attr(*, "dimnames")=List of 2
    ..... s: chr [1:2] "infant_mortality" "log_gdppc"
##
##
    .. .. .. .. .. : chr "log_gdppc"
     .. .. - attr(*, "term.labels")= chr "log_gdppc"
    .. .. ..- attr(*, "order")= int 1
##
    .. .. ..- attr(*, "intercept")= int 1
##
##
    .. .. ..- attr(*, "response")= int 1
##
     ..... attr(*, ".Environment")=<environment: R_GlobalEnv>
    .. .. attr(*, "predvars")= language list(infant_mortality, log_gdppc)
##
    ..... attr(*, "dataClasses")= Named chr [1:2] "numeric" "numeric"
##
    ..... attr(*, "names")= chr [1:2] "infant_mortality" "log_gdppc"
##
##
    ..- attr(*, "na.action")=Class 'omit' Named int [1:31] 9 12 14 24 45 46 61 65 72 78 ...
    ..... attr(*, "names")= chr [1:31] "14" "17" "NA" "29" ...
```

```
## - attr(*, "class")= chr "lm"

m1$coefficients

## (Intercept) log_gdppc
## 139.31467 -13.14033

m1$coefficients['(Intercept)']

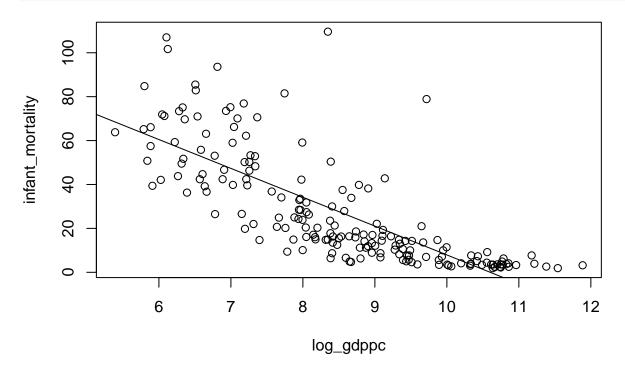
## (Intercept)
## 139.3147

m1$coefficients['log_gdppc']

## 1og_gdppc
## -13.14033
```

Now we can use them for other things, e.g plotting

```
plot(infant_mortality ~ log_gdppc, data = d_2010)
abline(a = m1$coefficients['(Intercept)'], b = m1$coefficients['log_gdppc'])
```



Report the model in a nice, journal-ready format

The stargazer library takes your model objects and generates tables in LaTeX. This package has a lot of customizing options, which you'll explore in the homework.

```
library(stargazer)
##
## Please cite as:
##
## Hlavac, Marek (2014). stargazer: LaTeX code and ASCII text for well-formatted regression and summar
## R package version 5.1. http://CRAN.R-project.org/package=stargazer
# LaTeX code that you can copy paste into LateX
stargazer(m1, m2)
##
## % Table created by stargazer v.5.1 by Marek Hlavac, Harvard University. E-mail: hlavac at fas.harvar
## % Date and time: Tue, Sep 15, 2015 - 01:13:26 PM
## \begin{table}[!htbp] \centering
##
    \caption{}
    \label{}
## \begin{tabular}{@{\extracolsep{5pt}}lcc}
## \\[-1.8ex]\hline
## \hline \\[-1.8ex]
## & \multicolumn{2}{c}{\textit{Dependent variable:}} \\
## \cline{2-3}
## \\[-1.8ex] & \multicolumn{2}{c}{infant\_mortality} \\
## \\[-1.8ex] & (1) & (2)\\
## \hline \\[-1.8ex]
## log\_gdppc & $-$13.140$^{***}$ & $-$10.585$^{***}$ \\
    & (0.791) & (0.982) \\
##
##
    & & \\
## number\_of\_physician & & $-$2.910^{***} \\
##
    & & (0.929) \\
##
    & & \\
## Constant & 139.315$^{***}$ & 121.107$^{***}$ \\
    & (6.804) & (7.431) \\
##
##
    & & \\
## \hline \\[-1.8ex]
## Observations & 186 & 143 \\
## R$^{2}$ & 0.600 & 0.708 \\
## Adjusted R$^{2}$ & 0.598 & 0.704 \\
## Residual Std. Error & 16.138 (df = 184) & 12.711 (df = 140) \\
## F Statistic & 276.047$^{***}$ (df = 1; 184) & 169.553$^{***}$ (df = 2; 140) \\
## \hline
## \hline \\[-1.8ex]
## \textit{Note:} & \multicolumn{2}{r}{$^{*}$p$<$0.1; $^{**}$p$<$0.05; $^{***}$p$<$0.01} \\
## \end{tabular}
## \end{table}
# If using knir, use the option results='asis'
stargazer(m1, m2)
```

[%] Table created by stargazer v.5.1 by Marek Hlavac, Harvard University. E-mail: hlavac at fas.harvard.edu % Date and time: Tue, Sep 15, 2015 - 01:13:26 PM

Table 1:

	(1)	(2)
log_gdppc	-13.140***	-10.585^{***}
	(0.791)	(0.982)
number_of_physician		-2.910***
		(0.929)
Constant	139.315***	121.107***
	(6.804)	(7.431)
Observations	186	143
\mathbb{R}^2	0.600	0.708
Adjusted R^2	0.598	0.704
Residual Std. Error	16.138 (df = 184)	12.711 (df = 140)
F Statistic	$276.047^{***} (df = 1; 184)$	$169.553^{***} (df = 2; 140)$
Note:	*n<0.1: **n<0.05: ***n<0.01	

Note:

*p<0.1; **p<0.05; ***p<0.01