Minimal R for Intro Stats

Randall Pruim, Project MOSAIC

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"Less volume, more creativity."

Mike McCarthy, Head Coach, Green Bay Packers

Mike McCarthy had signs proclaiming his "Less volume, more creativity" mantra hung on the office walls of all of his coordinators during one off-season. When asked about it, he said, "A lot of times you end up putting in a lot more volume, because you are teaching fundamentals and you are teaching concepts that you need to put in, but you may not necessarily use because they are building blocks for other concepts and variations that will come off of that ... In the offseason you have a chance to take a step back and tailor it more specifically towards your team and towards your players."

Statistics instructors using R face a similar dilemma. R is capable of so much that it is tempting to include this, and then that, and then the other, and then one more thing. Vectors and lists and recycling and coercion and functions and . . . It all seems so fundamental to the way R works. And when mastered, these concepts do become building blocks for other concepts and variations.

But when looking back at the end of a term, we have to admit that some of these things really aren't necessary to get the job done, and may do more harm than good for beginners. We too need to take a step back and tailor things toward our students and their abilities and needs. The colored commands on the next page are sufficient for an Introductory Statistics course that includes ANOVA, regression, and resampling techniques. The others are optional extras. This is followed by a 1-page sampler showing usage examples for some of the functions.

Note: These pages are intended as a guide for instructors, not as a reference card for students. Although they may also be useful for students, they would need supplementing with additional details.

The list of functions we present are not the only sufficient set of functions, but they were carefully chosen to fit as much as possible into a small number of paradigms. In particular,

1. We make use of the "formula interface" whenever possible.

Students will need the formula interface to do regression an ANOVA. Since we are going to teach it anyway, we use formulas as consistently and as often as we can. In some cases, my colleagues and I have written new functions or expanded the use of existing functions to serve this end. These functions are available in the mosaic package and are indicated in the comments in our palette. Some of the data sets are from the mosaicData package.

2. We use lattice graphics.

R has three separate high level plotting libraries (base, lattice, and ggplot2). Each has its advantages, but we choose lattice because it uses the same formula interface used elsewhere and because it encourages students to things about disaggregating data according to the values of covariates by making this very easy to do.

Whether you use this list or some other list, we encourage you to make a complete list of the commands you want your students to learn over the course of a semester. Organize them by topic. Organize them again by syntactic structure. Ask yourself how they look as a whole. Have you chosen a set of functions that fit well together? And most importantly: What is your creativity to volume quotient?

Help

apropos() ? ?? example()

Basic Calculations

Basic calculation works like a calculator. var()

```
# basic ops: + - * / ^ ( )
log(); exp(); sqrt()
log10(); abs(); choose()
```

Formula Interface

The following syntax (often with some parts omitted) is used for graphical summaries, numerical summaries, and inference procedures.

For plots:

- y: is y-axis variable
- x: is x-axis variable
- z: conditioning variable (separate panels)
- groups: conditioning variable (overlaid graphs)

For other things:

'y $\sim x \mid z'$ can usually be read 'y is modeled by (or depends on) x differently for each z'.

See the sampler for examples.

Numerical Summaries

favstats()

tallv()

mean()

These functions have a formula interface to match plotting.

mosaic

mosaic

mosaic augmented

```
median()
             # mosaic augmented
sd()
             # mosaic augmented
             # mosaic augmented
diffmean()
             # mosaic
             # mosaic augmented
quantile()
prop()
             # mosaic
perc()
             # mosaic
rank()
IQR()
             # mosaic augmented
```

min(); max() # mosaic augmented

Graphics (mostly lattice)

```
bwplot()
xyplot()
histogram() # mosaic augmented
densityplot()
freqpolygon() # mosaic
ggmath()
makeFun()
            # mosaic
plotFun()
            # mosaic
ladd()
            # mosaic
dotPlot()
            # mosaic
bargraph()
            # mosaic
xqqmath()
            # mosaic
mplot(data=HELPrct, 'scatter')
```

mplot(data=HELPrct, 'boxplot')

mplot(data=HELPrct, 'histogram')

Randomization/Simulation

```
rflip()  # mosaic
do()  # mosaic
sample()  # mosaic augmented
resample()  # with replacement
shuffle()  # mosaic
rbinom()
rnorm()  # etc, if needed
```

Distributions

```
pbinom(); pnorm();
xpnorm()  # mosaic augmented
pchisq(); pt()
qbinom(); qnorm();
qchisq(); qt()
plotDist()  # mosaic
```

Inference

```
# mosaic augmented
t.test()
binom.test() # mosaic augmented
prop.test()
             # mosaic augmented
xchisq.test() # mosaic
fisher.test()
pval()
              # mosaic
model <- lm() # linear models
summary(model)
coef(model)
confint(model) # mosaic augmented
anova(model)
makeFun(model) # mosaic
resid(model); fitted(model)
mplot(model) # mosaic
```

```
mplot(TukeyHSD(model))
model <- glm() # logistic reg.</pre>
```

Data

```
require(mosaicData) # load package
read.file() # mosaic
nrow(); ncol(); dim()
summary()
str()
names()
head(); tail()
with()
factor()
```

```
ntiles() # mosaic
cut()
c()
cbind(); rbind()
colnames()
rownames()
relevel()
reorder()
```

```
rep()
seq()
sort()
rank()
```

Data Transformation

```
select()
               # dplvr
mutate()
               # dplyr
filter()
               # dplvr
arrange()
               # dplyr
summarise()
               # dplyr
group_by()
               # dplyr
left_join()
               # dplyr
inner_join()
               # dplyr
merge()
```

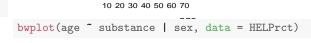
R Sampler for Intro Stats

```
rflip(6)
Flipping 6 coins [ Prob(Heads) = 0.5 ] ...
THTHHT
Number of Heads: 3 [Proportion Heads: 0.5]
do(2) * rflip(6)
  n heads tails
                     prop
              1 0.8333333
              2 0.6666667
coins <- do(1000) * rflip(6)</pre>
tally("heads, data = coins)
    1 2 3 4
                      5
 14 101 212 322 235 96 20
tally("heads, data = coins, format = "perc")
 1.4 10.1 21.2 32.2 23.5 9.6 2.0
tally("(heads >= 5 | heads <= 1), data = coins)
 TRUE FALSE
  231 769
```

```
histogram(~heads, data = coins, width = 1,
    groups = (heads >= 5 \mid heads <= 1))
```

```
0.3
0.2
0.1
                      heads
```

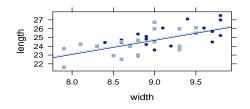
```
require(mosaicData) # this package contains data se pval(binom.test(~sex, data = HELPrct))
tally(sex ~ substance, data = HELPrct)
        substance
         alcohol cocaine heroin
sex
  female
              36
                      41
             141
                     111
                              94
  male
mean(age ~ sex, data = HELPrct)
  female
             male
36.25234 35.46821
diffmean(age ~ sex, data = HELPrct)
  diffmean
-0.7841284
favstats(age ~ sex, data = HELPrct)
     sex min Q1 median
                         Q3 max
1 female 21 31
                    35 40.5 58
2 male 19 30
                    35 40.0 60
      mean
                 sd n missing
1 36.25234 7.584858 107
2 35.46821 7.750110 346
densityplot(~age | sex, groups = substance,
    data = HELPrct, auto.key = TRUE)
                      alcohol
                      cocaine
                      heroin
                            10 20 30 40 50 60 70
```



```
female
                                 male
60
40
                                          •
30
     alcoholcocaine heroin alcoholcocaine heroin
```

```
p.value
1.931901e-30
confint(t.test(~age, data = HELPrct))
  mean of x lower
                      upper level
1 35.65342 34.9415 36.36534 0.95
model <- lm(length ~ width + sex,
```

```
data=KidsFeet)
ln <- makeFun(model)</pre>
ln( width=8.25, sex="B")
23.59689
xyplot(length ~ width, groups=sex,
       data=KidsFeet)
plotFun(ln(w, sex="B") ~ w,
        add=TRUE, col="skyblue")
plotFun(ln(w,sex="G") ~ w,
        add=TRUE, col="navy")
```



```
plotDist("chisq", df = 4)
```

