The three “back ticks” (”) must be followed by curly brackets “{”, and then “r” to tell the computer that you are using R code. This line is then closed off by another curly bracket “}”.

Anything before three more back ticks “”” are then considered R code (a script).

If any code in the document has just a backtick ‘ then nothing, then another backtick, then that word is just printed as if it were code, such as hey.

I’m reading in the bike lanes here.

```r
# readin is just a "label" for this code chunk
## code chunk is just a "chunk" of code, where this code usually
## does just one thing, aka a module
### comments are still # here
### you can do all your reading in there
### let's say we loaded some packages
library(stringr)
library(plyr)
library(dplyr)
```

```r
# Attaching package: 'dplyr'
#
# The following objects are masked from 'package:plyr':
#
# arrange, count, desc, failwith, id, mutate, rename, summarise,
# summarize
#
# The following objects are masked from 'package:stats':
#
# filter, lag
#
# The following objects are masked from 'package:base':
#
# intersect, setdiff, setequal, union
```

```r
fname <- "../data/Bike_Lanes.csv"
bike = read.csv(fname, as.is = TRUE)
```

You can write your introduction here.
Introduction

Bike lanes are in Baltimore. People like them. Why are they so long?

Exploratory Analysis

Let’s look at some plots of bike length. Let’s say we wanted to look at what affects bike length.

Plots of bike length

Note we made the subsection by using three “hashes” (pound signs): `###`. We can turn off R code output by using `echo = FALSE` on the knitr code chunk.
We have a total of 1505 rows.

What does it look like if we took the log (base 10) of the bike length:

```r
no.missyear$log.length <- log10(no.missyear$length)
### see here that if you specify the data argument, you don't need to do the $
boxplot(log.length ~ dateInstalled, data=no.missyear, main="Boxplots of Bike Lenght by Year", xlab="Year")
```
I want my boxplots colored, so I set the `col` argument.

```r
boxplot(log.length ~ dateInstalled, data=no.missyear, main="Boxplots of Bike Length by Year", xlab="Year", ylab="Bike Length", col="red")
```

As we can see, 2006 had a much higher bike length. What about for the type of bike path?
### type is a character, but when R sees a "character" in a "formula", then it automatically converts it
### a formula is something that has a y ~ x, which says I want to plot y against x
### or if it were a model you would do y ~ x, which meant regress against y

```r
boxplot(log.length ~ type, data=no.missyear, main="Boxplots of Bike Lenght by Year", xlab="Year", ylab="Bike Length")
```

What if we want to extract means by each type?

Let's show a few ways:

### tapply takes in vector 1, then does a function by vector 2, and then you tell what
### that function is

```r
tapply(no.missyear$log.length, no.missyear$type, mean)
```

```r
## Group.1 x
## 1 BIKE LANE 2.330611
## 2 CONTRAFLOW 2.087246
## 3 SHARED BUS BIKE 2.363005
## 4 SHARROW 2.256425
## 5 SIDEPATH 2.781829
## 6 SIGNED ROUTE 2.263746
```

### aggregate

```r
aggregate(x=no.missyear$log.length, by=list(no.missyear$type), FUN=mean)
```

```r
## Group.1  x
## 1 BIKE LANE 2.330611
## 2 CONTRAFLOW 2.087246
## 3 SHARED BUS BIKE 2.363005
## 4 SHARROW 2.256425
## 5 SIDEPATH 2.781829
## 6 SIGNED ROUTE 2.263746
```
### now let's specify the data argument and use a "formula" - much easier to read and
## more "intuitive"
aggregate(log.length ~ type, data=no.missyear, FUN=mean)

```r
## type log.length
## 1 BIKE LANE 2.330611
## 2 CONTRAFL ow 2.087246
## 3 SHARED BUS BIKE 2.363005
## 4 SHARROW 2.256425
## 5 SIDEPATH 2.781829
## 6 SIGNED ROUTE 2.263746
```

## ddply is from the plyr package
## takes in a data frame, (the first d refers to data.frame)
## splits it up by some variables (let's say type)
## then we'll use summarise to summarize whatever we want
## then returns a data.frame (the second d) - hence why it's ddply
## if we wanted to do it on a "list" thne return data.frame, it'd be ldply
```r
ddply(no.missyear, .(type), plyr::summarise,
       mean=mean(log.length))
```

```r
## Source: local data frame [6 x 2]
##
## type mean
## 1 BIKE LANE 2.330611
## 2 CONTRAFL ow 2.087246
## 3 SHARED BUS BIKE 2.363005
## 4 SHARROW 2.256425
## 5 SIDEPATH 2.781829
## 6 SIGNED ROUTE 2.263746
```

```r
no.missyear %>%
  group_by(type) %>%
  dplyr::summarise(mean=mean(log.length))
```

```r
## Source: local data frame [6 x 2]
##
## type mean
## 1 BIKE LANE 2.330611
## 2 CONTRAFL ow 2.087246
## 3 SHARED BUS BIKE 2.363005
## 4 SHARROW 2.256425
## 5 SIDEPATH 2.781829
## 6 SIGNED ROUTE 2.263746
```

ddply (and other functions in the plyr package) is cool because you can do multiple functions really easy.

Let's show a what if we wanted to go over type and dateInstalled:

```r
## For going over 2 variables, we need to do it over a "list" of vectors
tapply(no.missyear$log.length,
       list(no.missyear$type, no.missyear$dateInstalled),
       mean)
```
<table>
<thead>
<tr>
<th></th>
<th>2006</th>
<th>2007</th>
<th>2008</th>
<th>2009</th>
<th>2010</th>
<th>2011</th>
</tr>
</thead>
<tbody>
<tr>
<td>BIKE LANE</td>
<td>3.046261</td>
<td>2.351256</td>
<td>2.365728</td>
<td>2.381418</td>
<td>2.306994</td>
<td>2.242132</td>
</tr>
<tr>
<td>CONTRAFLOW</td>
<td>NA</td>
<td>NA</td>
<td>NA</td>
<td>NA</td>
<td>2.087246</td>
<td>NA</td>
</tr>
<tr>
<td>SHARED BUS BIKE</td>
<td>NA</td>
<td>NA</td>
<td>NA</td>
<td>2.350759</td>
<td>2.403824</td>
<td>NA</td>
</tr>
<tr>
<td>SHARROW</td>
<td>2.300954</td>
<td>2.220850</td>
<td>2.691814</td>
<td>2.247131</td>
<td>NA</td>
<td></td>
</tr>
<tr>
<td>SIDEPATH</td>
<td>NA</td>
<td>NA</td>
<td>2.625486</td>
<td>NA</td>
<td>2.773850</td>
<td>3.266816</td>
</tr>
<tr>
<td>SIGNED ROUTE</td>
<td>NA</td>
<td>2.287593</td>
<td>NA</td>
<td>NA</td>
<td>2.239475</td>
<td>2.210112</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th></th>
<th>2012</th>
<th>2013</th>
</tr>
</thead>
<tbody>
<tr>
<td>BIKE LANE</td>
<td>2.36151</td>
<td>2.408306</td>
</tr>
<tr>
<td>CONTRAFLOW</td>
<td>NA</td>
<td>NA</td>
</tr>
<tr>
<td>SHARED BUS BIKE</td>
<td>NA</td>
<td>NA</td>
</tr>
<tr>
<td>SHARROW</td>
<td>2.23636</td>
<td>NA</td>
</tr>
<tr>
<td>SIDEPATH</td>
<td>NA</td>
<td>NA</td>
</tr>
<tr>
<td>SIGNED ROUTE</td>
<td>NA</td>
<td>NA</td>
</tr>
</tbody>
</table>

```r
tapply(no.missyear$log.length, 
  list(no.missyear$type, no.missyear$dateInstalled), 
  mean, na.rm=TRUE)
```

```r
aggregate(log.length ~ type + dateInstalled, data=no.missyear, FUN=mean)
```

```r
aggregate(log.length ~ dateInstalled + log.length, data=no.missyear, FUN=mean)
```
OK let's do an linear model

```r
## type is a character, but when R sees a "character" in a "formula", then it automatically converts it into a factor
## a formula is something that has a y ~ x, which says I want to plot y against x
## or if it were a model you would do y ~ x, which meant regress against y
mod.type = lm(log.length ~ type, data=no.missyear)
mod.yr = lm(log.length ~ factor(dateInstalled), data=no.missyear)
mod.yrtype = lm(log.length ~ type + factor(dateInstalled), data=no.missyear)
summary(mod.type)
```

```r
## Call:
## lm(formula = log.length ~ type, data = no.missyear)
```
## Residuals:

<table>
<thead>
<tr>
<th>Min</th>
<th>1Q</th>
<th>Median</th>
<th>3Q</th>
<th>Max</th>
</tr>
</thead>
<tbody>
<tr>
<td>-1.51498</td>
<td>-0.19062</td>
<td>0.02915</td>
<td>0.23220</td>
<td>1.31021</td>
</tr>
</tbody>
</table>

## Coefficients:

| Estimate  | Std. Error | t value | Pr(>|t|) |
|-----------|------------|---------|----------|
| (Intercept) 2.33061 | 0.01487 | 156.703 | < 2e-16 *** |
| typeCONTRAFLOW -0.24337 | 0.10288 | -2.366 | 0.018127 * |
| typeSHARED BUS BIKE 0.03239 | 0.06062 | 0.534 | 0.593194 |
| typeSHARROW -0.07419 | 0.02129 | -3.484 | 0.000509 *** |
| typeSIDEPATH 0.45122 | 0.15058 | 2.997 | 0.002775 ** |
| typeSIGNED ROUTE -0.06687 | 0.02726 | -2.453 | 0.014300 * |

---

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 0.367 on 1499 degrees of freedom
Multiple R-squared: 0.01956, Adjusted R-squared: 0.01629
F-statistic: 5.98 on 5 and 1499 DF, p-value: 1.74e-05

That's rather UGLY, so let's use a package called `xtable` and then make this model into an `xtable` object and then print it out nicely.

```r
## Loading required package: xtable

# smod <- summary(mod.yr)
xtab <- xtable(mod.yr)

Well `xtable` can make html tables, so let's print this. We must tell R that the results is actually an html output, so we say the results should be embedded in the html “asis” (aka just print out whatever R spits out).

```r
tab <- xtable(mod.yr)
print.xtable(xtab, type="html")
```

Estimate Std. Error t value Pr(>|t|)
(Intercept) 3.0463 0.2600 11.71 0.0000
factor(dateInstalled)2007
factor(dateInstalled)2008
-0.7808
0.2613
-2.99
0.0029
factor(dateInstalled)2009
-0.6394
0.2631
-2.43
0.0152
factor(dateInstalled)2010
-0.7791
0.2605
-2.99
0.0028
factor(dateInstalled)2011
-0.8022
0.2626
-3.05
0.0023
factor(dateInstalled)2012
-0.7152
0.2625
-2.72
0.0065
factor(dateInstalled)2013
-0.6380
0.2849
-2.24
0.0253

OK, that’s pretty good, but let’s say we have all three models. Another package called stargazer can put models together easily and print them out. So xtable is really good when you are trying to print out a table (in html, otherwise make the table and use write.csv to get it in Excel and then format) really quickly and in a report. But it doesn’t work so well with many models together. So let’s use stargazer. Again, you need to use install.packages("stargazer") if you don’t have function.
OK, so what’s the difference here? First off, we said results are “markup”, so that it will not try to reformat the output. Also, I didn’t want those # for comments, so I just made comment an empty string “”.

```r
stargazer(mod.yr, mod.type, mod.yrtype, type="text")
```

<p>| | | |</p>
<table>
<thead>
<tr>
<th></th>
<th></th>
<th></th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Dependent variable:</strong></td>
<td><strong>log.length</strong></td>
<td></td>
</tr>
<tr>
<td></td>
<td>(1)</td>
<td>(2)</td>
</tr>
<tr>
<td>factor(dateInstalled)2007</td>
<td>-0.733***</td>
<td>-0.690***</td>
</tr>
<tr>
<td></td>
<td>(0.261)</td>
<td>(0.259)</td>
</tr>
<tr>
<td>factor(dateInstalled)2008</td>
<td>-0.781***</td>
<td>-0.742***</td>
</tr>
<tr>
<td></td>
<td>(0.261)</td>
<td>(0.260)</td>
</tr>
<tr>
<td>factor(dateInstalled)2009</td>
<td>-0.639**</td>
<td>-0.619**</td>
</tr>
<tr>
<td></td>
<td>(0.263)</td>
<td>(0.262)</td>
</tr>
<tr>
<td>factor(dateInstalled)2010</td>
<td>-0.779***</td>
<td>-0.736***</td>
</tr>
<tr>
<td></td>
<td>(0.260)</td>
<td>(0.259)</td>
</tr>
<tr>
<td>factor(dateInstalled)2011</td>
<td>-0.802***</td>
<td>-0.790***</td>
</tr>
<tr>
<td></td>
<td>(0.263)</td>
<td>(0.261)</td>
</tr>
<tr>
<td>factor(dateInstalled)2012</td>
<td>-0.715***</td>
<td>-0.700***</td>
</tr>
<tr>
<td></td>
<td>(0.262)</td>
<td>(0.261)</td>
</tr>
<tr>
<td>factor(dateInstalled)2013</td>
<td>-0.638**</td>
<td>-0.638**</td>
</tr>
<tr>
<td></td>
<td>(0.285)</td>
<td>(0.283)</td>
</tr>
<tr>
<td>typeCONTRAFLOW</td>
<td>-0.243**</td>
<td>-0.224**</td>
</tr>
<tr>
<td></td>
<td>(0.103)</td>
<td>(0.103)</td>
</tr>
<tr>
<td>typeSHARED BUS BIKE</td>
<td>0.032</td>
<td>-0.037</td>
</tr>
<tr>
<td></td>
<td>(0.061)</td>
<td>(0.069)</td>
</tr>
<tr>
<td>typeSHARROW</td>
<td>-0.074***</td>
<td>-0.064***</td>
</tr>
<tr>
<td></td>
<td>(0.021)</td>
<td>(0.023)</td>
</tr>
<tr>
<td>typeSIDEPATH</td>
<td>0.451***</td>
<td>0.483***</td>
</tr>
</tbody>
</table>
If we use

\texttt{stargazer(mod.yr, mod.type, mod.yrtype, type="html")}

Dependent variable:
log.length

(1)

(2)

(3)

factor(dateInstalled)2007
-0.733***
-0.690***
(0.261)
(0.259)

factor(dateInstalled)2008
-0.781***
-0.742***
(0.261)
(0.260)

factor(dateInstalled)2009
-0.639**
-0.619**
(0.263)
(0.262)

factor(dateInstalled)2010
-0.779***
<table>
<thead>
<tr>
<th>Date Installed</th>
<th>Type</th>
<th>Effect Size</th>
<th>p-value</th>
</tr>
</thead>
<tbody>
<tr>
<td>2011</td>
<td>CONTRAFLOW</td>
<td>-0.736***</td>
<td>(0.260)</td>
</tr>
<tr>
<td></td>
<td>SHARED BUS BIKE</td>
<td>0.032</td>
<td>(0.061)</td>
</tr>
<tr>
<td>2012</td>
<td>CONTRAFLOW</td>
<td>-0.802***</td>
<td>(0.259)</td>
</tr>
<tr>
<td></td>
<td>SHARED BUS BIKE</td>
<td>-0.074***</td>
<td>(0.069)</td>
</tr>
<tr>
<td>2013</td>
<td>CONTRAFLOW</td>
<td>-0.790***</td>
<td>(0.263)</td>
</tr>
<tr>
<td></td>
<td>SHARED BUS BIKE</td>
<td>-0.064***</td>
<td>(0.021)</td>
</tr>
<tr>
<td></td>
<td>SIDEPATH</td>
<td>0.451***</td>
<td>(0.023)</td>
</tr>
<tr>
<td></td>
<td>SHARROW</td>
<td>0.483***</td>
<td>(0.023)</td>
</tr>
</tbody>
</table>
type SIGNED ROUTE

\[-0.067^{**}\]
\[-0.067^{**}\]
\[(0.027)\]
\[(0.029)\]

Constant
\[3.046^{***}\]
\[2.331^{***}\]
\[3.046^{***}\]
\[(0.260)\]
\[(0.015)\]
\[(0.258)\]

Observations
1,505
1,505
1,505

R2
0.017
0.020
0.033

Adjusted R2
0.012
0.016
0.026

Residual Std. Error
0.368 (df = 1497)
0.367 (df = 1499)
0.365 (df = 1492)

F Statistic
3.691*** (df = 7; 1497)
5.980*** (df = 5; 1499)
4.285*** (df = 12; 1492)

Note:
\[p<0.1; \ p<0.05; \ p<0.01\]
Data Extraction

Let’s say I want to get data INTO my text. Like there are N number of bike lanes with a date installed that isn’t zero. There are 1505 bike lanes with a date installed after 2006. So you use one backtick ‘ and then you say “r” to tell that it’s R code. And then you run R code that gets evaluated and then returns the value. Let’s say you want to compute a bunch of things:

```r
### let's get number of bike lanes installed by year
n.lanes = ddply(no.missyear, .(dateInstalled), nrow)
names(n.lanes) <- c("date", "nlanes")
n2009 <- n.lanes$nlanes[ n.lanes$date == 2009]
n2010 <- n.lanes$nlanes[ n.lanes$date == 2010]
getwd()
```

Now I can just say there are 86 lanes in 2009 and 625 in 2010.

```r
fname <- "../../data/Charm_City_Circulator_Ridership.csv"
# fname <- file.path(data.dir, "Charm_City_Circulator_Ridership.csv")
## file.path takes a directory and makes a full name with a full file path
charm = read.csv(fname, as.is=TRUE)
library(chron)
days = levels(weekdays(1, abbreviate=FALSE))
charm$day <- factor(charm$day, levels=days)
charm$date <- as.Date(charm$date, format="%m/%d/%Y")
cn <- colnames(charm)
daily <- charm[, c("day", "date", "daily")]
```

```r
charm$daily <- NULL
require(reshape)
## Loading required package: reshape
##
## Attaching package: 'reshape'
##
## The following object is masked from 'package:dplyr':
##
## rename
##
## The following objects are masked from 'package:plyr':
##
## rename, round_any
long.charm <- melt(charm, id.vars = c("day", "date"))
long.charm$type[ grepl("Alightings", long.charm$variable)] <- "Alightings"
long.charm$type[ grepl("Average", long.charm$variable)] <- "Average"
long.charm$line[ grepl("purple", long.charm$variable)] <- "purple"
```
long.charm$line[grepl("green", long.charm$variable)] <- "green"
long.charm$line[grepl("banner", long.charm$variable)] <- "banner"
long.charm$variable <- NULL
long.charm$line <- factor(long.charm$line, levels=c("orange", "purple", "green", "banner"))
head(long.charm)

##  day date   value  type line
## 1 Monday 2010-01-11 877 Boardings orange
## 2 Tuesday 2010-01-12 777 Boardings orange
## 3 Wednesday 2010-01-13 1203 Boardings orange
## 4 Thursday 2010-01-14 1194 Boardings orange
## 5 Friday 2010-01-15 1645 Boardings orange
## 6 Saturday 2010-01-16 1457 Boardings orange

### NOW R has a column of day, the date, a "value", the type of value and the circulator line that corresponds to it
### value is now either the Alightings, Boardings, or Average from the charm dataset

Let's do some plotting now!

require(ggplot2)

## Loading required package: ggplot2

## Warning: package 'ggplot2' was built under R version 3.2.3

### let's make a "ggplot"
### the format is ggplot(dataframe, aes(x=COLNAME, y=COLNAME))
### where COLNAME are colnames of the dataframe
### you can also set color to a different factor
### other options in AES (fill, alpha level -which is the "transparency" of points)
g <- ggplot(long.charm, aes(x=date, y=value, color=line))
### let's change the colors to what we want- doing this manually, not letting it choose
### for me
g <- g + scale_color_manual(values=c("orange", "purple", "green", "blue"))
### plotting points
g + geom_point()

## Warning: Removed 5328 rows containing missing values (geom_point).
### Let's make Lines!

```r
g + geom_line()
```

## Warning: Removed 5043 rows containing missing values (geom_path).
```r
### let's make a new plot of points

```gpoint <- g + geom_point()
```n
### let's plot the value by the type of value - boardings/average, etc

gpoint + facet_wrap(~ type)

## Warning: Removed 5328 rows containing missing values (geom_point).

```
OK let's turn off some warnings - making `warning=FALSE` (in knitr) as an option.

```r
## let's compare vertically
p + facet_wrap(~ type, ncol=1)
```
We can also smooth the data to give us an overall idea of how the average changes over time. I don’t want to do a standard error (se).

```
## let’s smooth this - get a rough estimate of what’s going on
gfacet + geom_smooth(se=FALSE)
```
OK, I’ve seen enough code, let’s turn that off, using `echo=FALSE`.

There are still messages, but we can turn these off with `message = FALSE`