## Knitr

## Introduction to R for Public Health Researchers

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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.

```
# 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(dplyr)
library(tidyr)
fname <- "http://www.aejaffe.com/summerR_2016/data/Bike_Lanes.csv"
bike = read.csv(fname, as.is = TRUE)</pre>
```

You can write your introduction here.

## Introduction

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

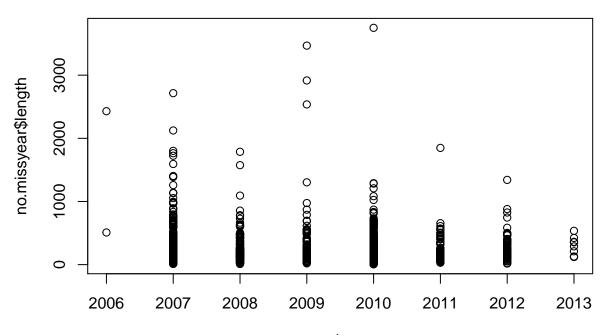
### **Exploratory Analysis**

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

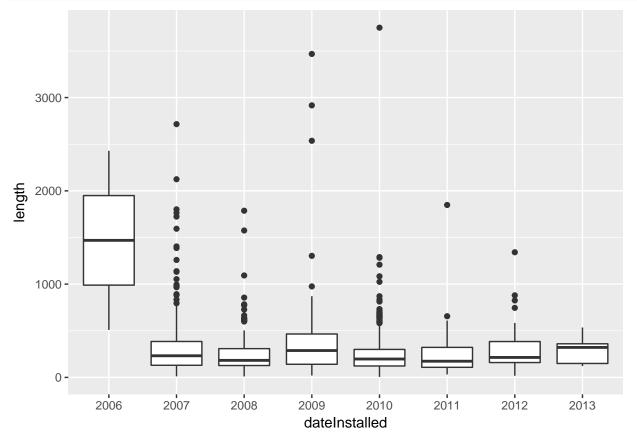
Note we made the subsection by using three "hashes" (pound signs): ###.

We can turn off R code output by using echo = FALSE on the knitr code chunks.



## no.missyear\$dateInstalled

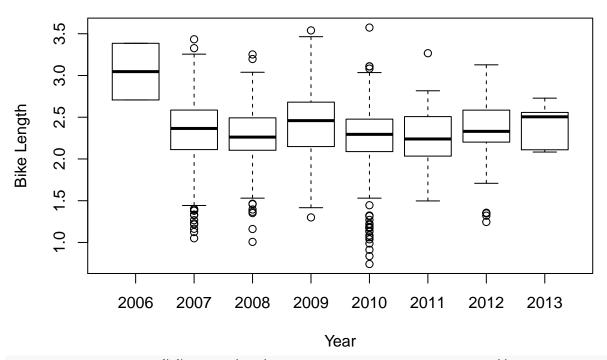
```
no.missyear = no.missyear %>% mutate(dateInstalled = factor(dateInstalled))
library(ggplot2)
gbox = no.missyear %>% ggplot(aes(x = dateInstalled, y = length)) + geom_boxplot()
print(gbox)
```



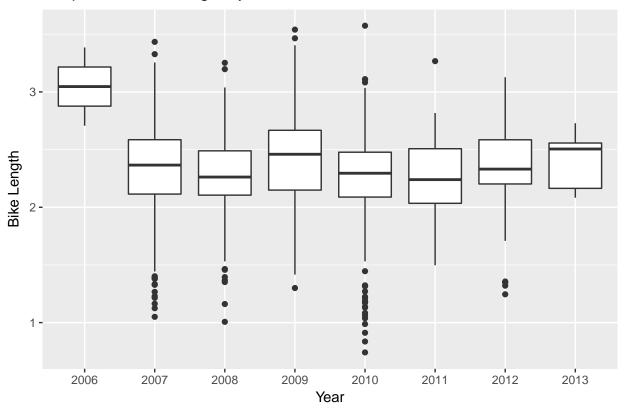
We have a total of 1505 rows.

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

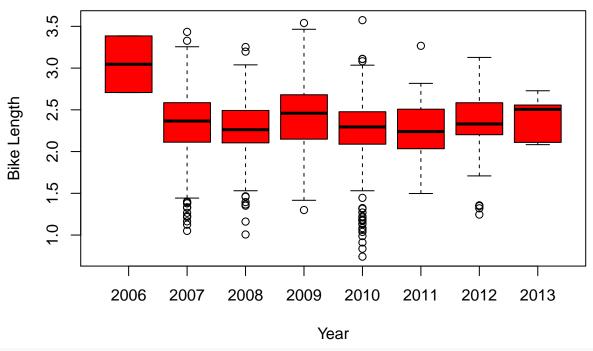
## **Boxplots of Bike Lenght by Year**



```
glogbox = no.missyear %>% ggplot(aes(x = dateInstalled, y = log.length)) + geom_boxplot() +
    ggtitle("Boxplots of Bike Lenght by Year") +
    xlab("Year") +
    ylab("Bike Length")
print(glogbox)
```

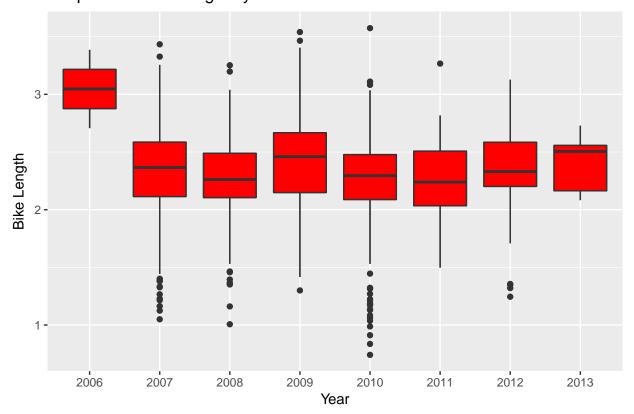


I want my boxplots colored, so I set the col argument.



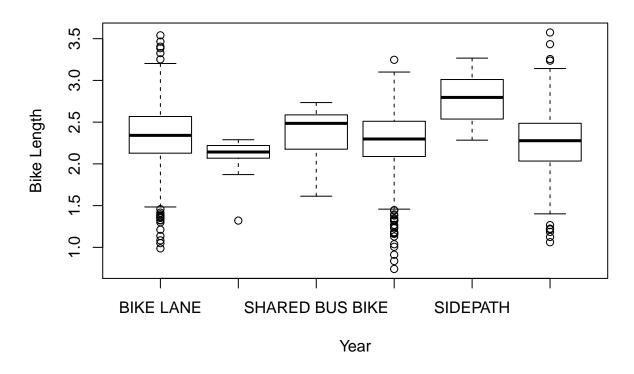
glogbox + geom\_boxplot(fill = "red")

# Boxplots of Bike Lenght by Year



As we can see, 2006 had a much higher bike length. What about for the type of bike path?

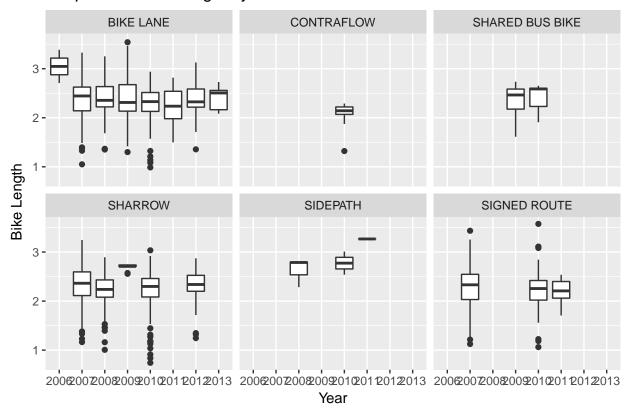
```
### type is a character, but when R sees a "character" in a "formula", then it automatically converts i
### a formula is something that has a y ~ x, which says I want to plot y against x
### or if it were a model you would do y ~ x, which meant regress against y
boxplot(log.length ~ type, data=no.missyear, main="Boxplots of Bike Length by Year", xlab="Year", ylab=
```



## **Multiple Facets**

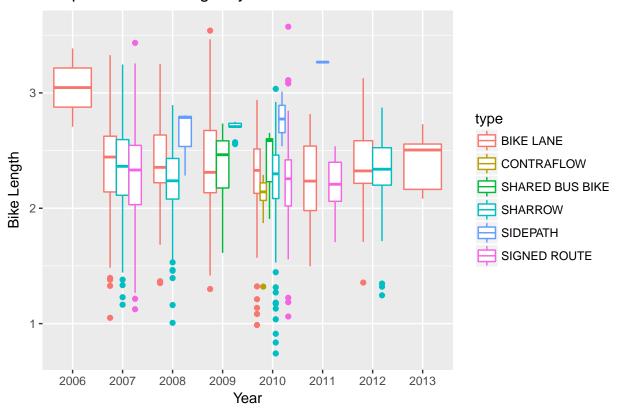
We can do the plot with different panels for each type.

```
glogbox + facet_wrap(~ type)
```



NOTE, this is different than if we colored on type:

glogbox + aes(colour = type)



#### Means by type

What if we want to extract means by each type?

Let's show a few ways:

```
no.missyear %>% group_by(type) %>%
  dplyr::summarise(mean = mean(log.length))
## # A tibble: 6 × 2
##
                 type
                          mean
##
                <chr>
                         <dbl>
## 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
Let's show a what if we wanted to go over type and dateInstalled:
```

```
no.missyear %>% group_by(type, dateInstalled) %>%
  dplyr::summarise(mean = mean(log.length),
        median = median(log.length),
        Std.Dev = sd(log.length))
```

```
## Source: local data frame [22 x 5]
## Groups: type [?]
```

```
##
##
                 type dateInstalled
                                                median
                                                         Std.Dev
                                         mean
##
                <chr>>
                                        <dbl>
                                                 <dbl>
## 1
            BIKE LANE
                                2006 3.046261 3.046261 0.4797354
## 2
            BIKE LANE
                                2007 2.351256 2.444042 0.4066225
                               2008 2.365728 2.354641 0.3891624
## 3
            BIKE LANE
                               2009 2.381418 2.311393 0.4944744
            BIKE LANE
## 5
            BIKE LANE
                               2010 2.306994 2.328486 0.3207591
## 6
            BIKE LANE
                               2011 2.242132 2.235462 0.3339777
## 7
            BIKE LANE
                               2012 2.361510 2.323863 0.2852810
## 8
            BIKE LANE
                                2013 2.408306 2.505012 0.2404060
                                2010 2.087246 2.142250 0.2565511
## 9
           CONTRAFLOW
## 10 SHARED BUS BIKE
                                2009 2.350759 2.463997 0.3060951
## # ... with 12 more rows
```

#### Linear Models

```
OK let's do some linear model
### type is a character, but when R sees a "character" in a "formula", then it automatically converts i
### 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 \sim 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)
##
## lm(formula = log.length ~ type, data = no.missyear)
## Residuals:
                  1Q
                      Median
                                    3Q
                                            Max
## -1.51498 -0.19062 0.02915 0.23220
                                       1.31021
## Coefficients:
##
                       Estimate Std. Error t value Pr(>|t|)
                                   0.01487 156.703 < 2e-16 ***
## (Intercept)
                       2.33061
## 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 '**' 0.05 '.' 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 pander and then make this model into an pander object and then print it out nicely.

## Grabbing coefficients

We can use the coef function on a summary, or do smod\$coef to get the coefficients. But they are in a matrix:

```
smod = summary(mod.type)
coef(smod)
##
                         Estimate Std. Error
                                                 t value
                                                              Pr(>|t|)
## (Intercept)
                       2.33061129 0.01487281 156.7027729 0.0000000000
## typeCONTRAFLOW
                      -0.24336564 0.10287662 -2.3656069 0.0181272020
## typeSHARED BUS BIKE 0.03239334 0.06062453 0.5343274 0.5931943055
## typeSHARROW
                      -0.07418617 0.02129463 -3.4837969 0.0005085795
## typeSIDEPATH
                       0.45121749 0.15057577 2.9966142 0.0027748128
## typeSIGNED ROUTE
                      -0.06686556 0.02726421 -2.4525034 0.0142999055
class(coef(smod))
## [1] "matrix"
```

#### Broom package

The broom package can "tidy" up the output to actually put the terms into a column of a data.frame that you can grab values from:

```
library(broom)
smod2 = tidy(mod.type)
class(smod2)
## [1] "data.frame"
better = smod2 %>% mutate(term = str_replace(term, "^type", ""))
better
##
                        estimate std.error
                                              statistic
                                                             p.value
                term
## 1
                     2.33061129 0.01487281 156.7027729 0.0000000000
         (Intercept)
## 2
         CONTRAFLOW -0.24336564 0.10287662 -2.3656069 0.0181272020
## 3 SHARED BUS BIKE 0.03239334 0.06062453 0.5343274 0.5931943055
## 4
             SHARROW -0.07418617 0.02129463 -3.4837969 0.0005085795
## 5
            SIDEPATH 0.45121749 0.15057577
                                              2.9966142 0.0027748128
## 6
        SIGNED ROUTE -0.06686556 0.02726421 -2.4525034 0.0142999055
better %>% filter(term == "SIDEPATH")
         term estimate std.error statistic
                                                p.value
## 1 SIDEPATH 0.4512175 0.1505758 2.996614 0.002774813
write.csv(better, file = "Best_Model_Coefficients.csv")
BUT I NEEEEEED an XLSX! The xlsx package can do it, but I still tend to use CSVs.
library(xlsx)
```

### Testing Nested Models

The anova command will test nested models and give you a table of results:

write.xlsx(better, file = "Best\_Model\_Coefficients.xlsx")

```
my_lrtest = anova(mod.yrtype, mod.yr)
print(my_lrtest)
## Analysis of Variance Table
##
## Model 1: log.length ~ type + factor(dateInstalled)
## Model 2: log.length ~ factor(dateInstalled)
               RSS Df Sum of Sq
    Res.Df
## 1
       1492 199.10
       1497 202.47 -5
                        -3.3681 5.048 0.000136 ***
## 2
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
print(tidy(my_lrtest))
##
     res.df
                 rss df
                            sumsq statistic
                                                  p.value
## 1
       1492 199.0977 NA
                               NA
                                                       NA
## 2
       1497 202.4658 -5 -3.368136 5.048034 0.0001360178
Similarly with year:
my_lrtest = anova(mod.yrtype, mod.type)
print(tidy(my_lrtest))
##
     res.df
                 rss df
                            sumsq statistic
                                                 p.value
## 1
       1492 199.0977 NA
                               NΑ
                                          NA
                                                      NΑ
       1499 201.9321 -7 -2.834384
                                   3.034333 0.003588298
```

## Pander

Pander can output tables (as well as other things such as models), so let's print this using the pander command from the pander package. So pander 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.

```
# devtools::install_github('Rapporter/pander') # need this version!
library(pander)
pander(mod.yr)
```

Table 1: Fitting linear model: log.length ~ factor(dateInstalled)

	Estimate	Std. Error	t value	$\Pr(> t )$
(Intercept)	3.046	0.26	11.71	2.181e-30
factor(dateInstalled)2007	-0.7332	0.2608	-2.812	0.004987
factor(dateInstalled)2008	-0.7808	0.2613	-2.988	0.002852
factor(dateInstalled)2009	-0.6394	0.2631	-2.431	0.01518
factor(dateInstalled)2010	-0.7791	0.2605	-2.991	0.002825
factor(dateInstalled)2011	-0.8022	0.2626	-3.055	0.002292
factor(dateInstalled)2012	-0.7152	0.2625	-2.725	0.006509
factor(dateInstalled)2013	-0.638	0.2849	-2.239	0.02527

It is the same if we write out the summary, but more information is in the **footer**.

ASIDE: the aov function fits what you think of when you think ANOVA.

## pander(summary(mod.yr))

	Estimate	Std. Error	t value	$\Pr(> t )$
(Intercept)	3.046	0.26	11.71	2.181e-30
factor(dateInstalled)2007	-0.7332	0.2608	-2.812	0.004987
factor(date Installed) 2008	-0.7808	0.2613	-2.988	0.002852
factor(date Installed) 2009	-0.6394	0.2631	-2.431	0.01518
${\it factor(dateInstalled)2010}$	-0.7791	0.2605	-2.991	0.002825
${\it factor(dateInstalled)2011}$	-0.8022	0.2626	-3.055	0.002292
${\it factor(dateInstalled)2012}$	-0.7152	0.2625	-2.725	0.006509
${\it factor}({\it dateInstalled}) 2013$	-0.638	0.2849	-2.239	0.02527

Table 3: Fitting linear model: log.length ~ factor(dateInstalled)

Observations	Residual Std. Error	$R^2$	Adjusted $\mathbb{R}^2$
1505	0.3678	0.01697	0.01237

## **Formatting**

Let's format the rows and the column names a bit better:

## Changing the terms

```
ptable = tidy(mod.yr)
ptable$term = ptable$term %>%
    str_replace(fixed("factor(dateInstalled)"), "") %>%
    str_replace(fixed("(Intercept)"), "Intercept")
```

### Column Names

Now we can reset the column names if we didn't like them before:

```
colnames(ptable) = c("Variable", "Beta", "SE", "tstatistic", "p.value")
pander(ptable)
```

Variable	Beta	SE	tstatistic	p.value
Intercept	3.046	0.26	11.71	2.181e-30
2007	-0.7332	0.2608	-2.812	0.004987
2008	-0.7808	0.2613	-2.988	0.002852
2009	-0.6394	0.2631	-2.431	0.01518
2010	-0.7791	0.2605	-2.991	0.002825
2011	-0.8022	0.2626	-3.055	0.002292
2012	-0.7152	0.2625	-2.725	0.006509
2013	-0.638	0.2849	-2.239	0.02527

#### Confidence Intervals

Let's say we want the beta, the 95% CI. We can use confint on the model, merge it to ptable and then

paste the columns together (after rounding) with a comma and bound them in parentheses.

```
cint = confint(mod.yr)
print(cint)
##
                                 2.5 %
                                             97.5 %
## (Intercept)
                              2.536168 3.55635353
## factor(dateInstalled)2007 -1.244725 -0.22177042
## factor(dateInstalled)2008 -1.293400 -0.26827336
## factor(dateInstalled)2009 -1.155435 -0.12345504
## factor(dateInstalled)2010 -1.289978 -0.26816090
## factor(dateInstalled)2011 -1.317344 -0.28710724
## factor(dateInstalled)2012 -1.229999 -0.20032262
## factor(dateInstalled)2013 -1.196733 -0.07917559
print(class(cint))
## [1] "matrix"
Tidying it up
cint = tidy(cint)
colnames(cint) = c("Variable", "lower", "upper")
cint$Variable = cint$Variable %>%
  str_replace(fixed("factor(dateInstalled)"), "") %>%
  str_replace(fixed("(Intercept)"), "Intercept")
ptable = left_join(ptable, cint, by = "Variable")
ptable = ptable %>% mutate(lower = round(lower, 2),
                           upper = round(lower, 2),
                           Beta = round(Beta, 2),
                           p.value = ifelse(p.value < 0.01, "< 0.01",</pre>
                                            round(p.value,2)))
ptable = ptable %>% mutate(ci = paste0("(", lower, ", ", upper, ")"))
ptable = dplyr::select(ptable, Beta, ci, p.value)
pander(ptable)
```

Beta	ci	p.value
3.05	(2.54, 2.54)	< 0.01
-0.73	(-1.24, -1.24)	< 0.01
-0.78	(-1.29, -1.29)	< 0.01
-0.64	(-1.16, -1.16)	0.02
-0.78	(-1.29, -1.29)	< 0.01
-0.8	(-1.32, -1.32)	< 0.01
-0.72	(-1.23, -1.23)	< 0.01
-0.64	(-1.2, -1.2)	0.03

### Multiple Models

OK, that's pretty good, but let's say we have all three models. You can't put doesn't work so well with *many* models together.

```
# pander(mod.yr, mod.yrtype) does not work
# pander(list(mod.yr, mod.yrtype)) # will give 2 separate tables
```

If we use the memisc package, we can combine the models:

```
library(memisc)
mtab_all <- mtable("Model Year" = mod.yr,</pre>
                   "Model Type" = mod.type,
                   "Model Both" = mod.yrtype,
                   summary.stats = c("sigma", "R-squared", "F", "p", "N"))
print(mtab_all)
##
## Calls:
## Model Year: lm(formula = log.length ~ factor(dateInstalled), data = no.missyear)
## Model Type: lm(formula = log.length ~ type, data = no.missyear)
## Model Both: lm(formula = log.length ~ type + factor(dateInstalled), data = no.missyear)
##
                                        Model Year Model Type Model Both
##
     (Intercept)
                                        3.046***
                                                     2.331***
                                                                 3.046***
                                        (0.260)
                                                    (0.015)
                                                                (0.258)
##
     factor(dateInstalled): 2007/2006 -0.733**
##
                                                                -0.690**
##
                                        (0.261)
                                                                (0.259)
##
     factor(dateInstalled): 2008/2006 -0.781**
                                                                -0.742**
##
                                        (0.261)
                                                                (0.260)
##
     factor(dateInstalled): 2009/2006 -0.639*
                                                                -0.619*
##
                                        (0.263)
                                                                (0.262)
     factor(dateInstalled): 2010/2006 -0.779**
##
                                                                -0.736**
                                                                (0.259)
##
                                        (0.260)
##
     factor(dateInstalled): 2011/2006 -0.802**
                                                                -0.790**
##
                                                                (0.261)
                                        (0.263)
##
     factor(dateInstalled): 2012/2006 -0.715**
                                                                -0.700**
##
                                                                (0.261)
                                        (0.262)
     factor(dateInstalled): 2013/2006 -0.638*
##
                                                                -0.638*
##
                                        (0.285)
                                                                (0.283)
##
     type: CONTRAFLOW/BIKE LANE
                                                    -0.243*
                                                                -0.224*
##
                                                    (0.103)
                                                                (0.103)
##
     type: SHARED BUS BIKE/BIKE LANE
                                                     0.032
                                                                -0.037
##
                                                    (0.061)
                                                                 (0.069)
     type: SHARROW/BIKE LANE
##
                                                    -0.074***
                                                                -0.064**
##
                                                    (0.021)
                                                                (0.023)
     type: SIDEPATH/BIKE LANE
##
                                                     0.451**
                                                                 0.483**
##
                                                    (0.151)
                                                                 (0.150)
##
     type: SIGNED ROUTE/BIKE LANE
                                                    -0.067*
                                                                -0.067*
##
                                                    (0.027)
                                                                 (0.029)
##
##
     sigma
                                            0.4
                                                        0.4
                                                                     0.4
##
                                                        0.0
                                                                     0.0
    R-squared
                                            0.0
##
                                            3.7
                                                        6.0
                                                                     4.3
##
                                            0.0
                                                        0.0
                                                                     0.0
##
                                         1505
                                                     1505
                                                                 1505
```

If you want to write it out (for Excel), it is tab delimited:

```
write.mtable(mtab_all, file = "my_tab.txt")
```

	Model Year	Model Type	Model Both
(Intercept)	3.046***	2.331***	3.046***
	(0.260)	(0.015)	(0.258)
factor(dateInstalled): 2007/2006	-0.733**		-0.690**
	(0.261)		(0.259)
factor(dateInstalled): 2008/2006	-0.781**		-0.742**
	(0.261)		(0.260)
factor(dateInstalled): 2009/2006	-0.639*		-0.619*
	(0.263)		(0.262)
factor(dateInstalled): 2010/2006	-0.779**		-0.736**
	(0.260)		(0.259)
factor(dateInstalled): 2011/2006	-0.802**		-0.790**
	(0.263)		(0.261)
factor(dateInstalled): 2012/2006	-0.715**		-0.700**
	(0.262)		(0.261)
factor(dateInstalled): 2013/2006	-0.638*		-0.638*
	(0.285)		(0.283)
type: CONTRAFLOW/BIKE		-0.243*	-0.224*
$\mathbf{LANE}$		(0.103)	(0.103)
type: SHARED BUS BIKE/BIKE		0.032	-0.037
$\mathbf{LANE}$		(0.061)	(0.069)
type: SHARROW/BIKE LANE		-0.074***	-0.064**
		(0.021)	(0.023)
type: SIDEPATH/BIKE LANE		0.451**	0.483**
		(0.151)	(0.150)
type: SIGNED ROUTE/BIKE		-0.067*	-0.067*
$\mathbf{LANE}$		(0.027)	(0.029)
$\mathbf{sigma}$	0.4	0.4	0.4
R-squared	0.0	0.0	0.0
${f F}$	3.7	6.0	4.3
p	0.0	0.0	0.0
${f N}$	1505	1505	1505

## Not covered - making mtable better:

```
renamer = function(model) {
  names(model$coefficients) = names(model$coefficients) %>%
  str_replace(fixed("factor(dateInstalled)"), "") %>%
  str_replace(fixed("(Intercept)"), "Intercept")
  names(model$contrasts) = names(model$contrasts) %>%
  str_replace(fixed("factor(dateInstalled)"), "") %>%
  str_replace(fixed("(Intercept)"), "Intercept")
  return(model)
}
mod.yr = renamer(mod.yr)
mod.yrtype = renamer(mod.yrtype)
mod.type = renamer(mod.type)

mtab_all_better <- mtable("Model Year" = mod.yr,</pre>
```

	Model Year	Model Type	Model Both
Intercept	3.046***	2.331***	3.046***
	(0.260)	(0.015)	(0.258)
2007	-0.733***	, ,	-0.690**
	(0.261)		(0.259)
2008	-0.781**		-0.742**
	(0.261)		(0.260)
$\boldsymbol{2009}$	-0.639*		-0.619*
	(0.263)		(0.262)
2010	-0.779**		-0.736**
	(0.260)		(0.259)
2011	-0.802**		-0.790**
	(0.263)		(0.261)
$\boldsymbol{2012}$	-0.715***		-0.700**
	(0.262)		(0.261)
2013	-0.638*		-0.638*
	(0.285)		(0.283)
type: CONTRAFLOW/BIKE	, ,	-0.243*	-0.224*
LANE		(0.103)	(0.103)
type: SHARED BUS BIKE/BIKE		0.032	-0.037
LANE		(0.061)	(0.069)
type: SHARROW/BIKE LANE		-0.074***	-0.064**
,		(0.021)	(0.023)
type: SIDEPATH/BIKE LANE		0.451**	0.483**
,		(0.151)	(0.150)
type: SIGNED ROUTE/BIKE		-0.067*	-0.067*
LANE		(0.027)	(0.029)
sigma	0.4	0.4	0.4
R-squared	0.0	0.0	0.0
$\mathbf{F}$	3.7	6.0	4.3
p	0.0	0.0	0.0
$\hat{f N}$	1505	1505	1505

Another package called stargazer can put models together easily and print them out. So let's use stargazer. Again, you need to use install.packages("stargazer") if you don't have function.

## require(stargazer)

OK, so what's the difference here? First off, we said results are "markup", so that it will not try to reform at the output. Also, I didn't want those # for comments, so I just made comment an empty string "".

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

Dependent variable:

log.length

(1) (2) (3)

2007	 -0.733***		 -0.690***
2001	(0.261)		(0.259)
	(0.201)		(0.203)
2008	-0.781***		-0.742***
	(0.261)		(0.260)
2009	-0.639**		-0.619**
	(0.263)		(0.262)
2010	-0.779***		-0.736***
2010	(0.260)		(0.259)
	(0.200)		(0.203)
2011	-0.802***		-0.790***
	(0.263)		(0.261)
2012	-0.715***		-0.700***
	(0.262)		(0.261)
2013	-0.638**		-0.638**
2013	(0.285)		(0.283)
	(0.200)		(0.203)
typeCONTRAFLOW		-0.243**	-0.224**
•		(0.103)	(0.103)
typeSHARED BUS BIKE		0.032	-0.037
		(0.061)	(0.069)
typeSHARROW		-0.074***	-0.064***
oj posimino.		(0.021)	(0.023)
typeSIDEPATH		0.451***	0.483***
		(0.151)	(0.150)
typeSIGNED ROUTE		-0.067**	-0.067**
		(0.027)	(0.029)
Constant	3.046***	2.331***	3.046***
oons can c	(0.260)	(0.015)	(0.258)
	(**=**,	(3.32-3)	(**/
	4 505	4 505	4 505
Observations R2	1,505 0.017	1,505 0.020	1,505 0.033
Adjusted R2	0.017	0.020	0.035
		0.367 (df = 1499)	
F Statistic			4.285*** (df = 12; 1492)
			==========
Note:		*p	<0.1; **p<0.05; ***p<0.01

If we use

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

Dependent variable:

 $\log.length$ 

- (1)
- (2)
- (3)
- 2007
- -0.733\*\*\*
- -0.690\*\*\*
- (0.261)
- (0.259)
- 2008
- -0.781\*\*\*
- -0.742\*\*\*
- (0.261)
- (0.260)
- 2009
- -0.639\*\*
- -0.619\*\*
- (0.263)
- (0.262)
- 2010
- -0.779\*\*\*
- -0.736\*\*\*
- (0.260)
- (0.259)
- 2011
- -0.802\*\*\*
- -0.790\*\*\*
- (0.263)
- (0.261)
- 2012
- -0.715\*\*\*
- -0.700\*\*\*
- (0.262)
- (0.261)
- 2013
- -0.638\*\*
- -0.638\*\*

(0.285)
(0.283)
typeCONTRAFLOW
-0.243**
-0.224**
(0.103)
(0.103)
typeSHARED BUS BIKE
0.032
-0.037
(0.061)
(0.069)
${\it typeSHARROW}$
-0.074***
-0.064***
(0.021)
(0.023)
${\it type} {\it SIDEPATH}$
0.451***
0.483***
(0.151)
(0.150)
${\it type} {\it SIGNED} \ {\it ROUTE}$
-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 evaulated and then returns the value. Let's say you want to compute a bunch of things:

```
### let's get number of bike lanes installed by year
n.lanes = no.missyear %>% group_by(dateInstalled) %>% dplyr::summarize(n())
class(n.lanes)
## [1] "tbl_df"
                     "tbl"
                                   "data.frame"
print(n.lanes)
## # A tibble: 8 × 2
##
     dateInstalled `n()`
##
            <fctr> <int>
## 1
               2006
                        2
## 2
              2007
                      368
## 3
               2008
                      206
               2009
                       86
## 4
## 5
               2010
                      625
## 6
               2011
                      101
```

```
## 7
               2012
                      107
## 8
               2013
                       10
n.lanes = as.data.frame(n.lanes)
print(n.lanes)
     dateInstalled n()
## 1
              2006
## 2
              2007 368
## 3
              2008 206
## 4
               2009 86
## 5
               2010 625
## 6
               2011 101
## 7
               2012 107
## 8
               2013 10
colnames(n.lanes) <- c("date", "nlanes")</pre>
n2009 <- filter(n.lanes, date == 2009)
n2010 \leftarrow filter(n.lanes, date == 2010)
getwd()
## [1] "/Users/johnmuschelli/Dropbox/Teaching/winterR_2017/Knitr/lecture"
Now I can just say there are 2009, 86 lanes in 2009 and 2010, 625 in 2010.
fname <- "http://www.aejaffe.com/summerR_2016/data/Charm_City_Circulator_Ridership.csv"</pre>
## 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)</pre>
charm$date <- as.Date(charm$date, format="%m/%d/%Y")
cn <- colnames(charm)</pre>
daily <- charm[, c("day", "date", "daily")]</pre>
charm$daily <- NULL
require(reshape)
long.charm <- melt(charm, id.vars = c("day", "date"))</pre>
long.charm$type <- "Boardings"</pre>
long.charm$type[ grepl("Alightings", long.charm$variable)] <- "Alightings"</pre>
long.charm$type[ grepl("Average", long.charm$variable)] <- "Average"</pre>
long.charm$line <- "orange"</pre>
long.charm$line[ grepl("purple", long.charm$variable)] <- "purple"</pre>
long.charm$line[ grepl("green", long.charm$variable)] <- "green"</pre>
long.charm$line[ grepl("banner", long.charm$variable)] <- "banner"</pre>
long.charm$variable <- NULL</pre>
long.charm$line <-factor(long.charm$line, levels=c("orange", "purple",</pre>
                                                       "green", "banner"))
head(long.charm)
##
                      date value
                                        type
           day
## 1
        Monday 2010-01-11 877 Boardings orange
```

Tuesday 2010-01-12 777 Boardings orange

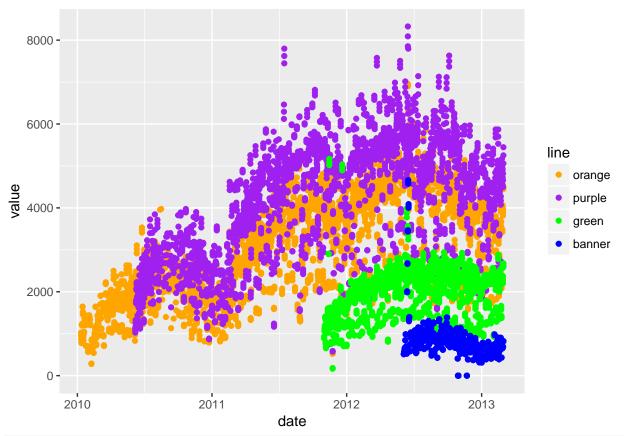
## 2

```
## 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
```

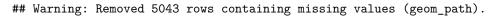
Let's do some plotting now!

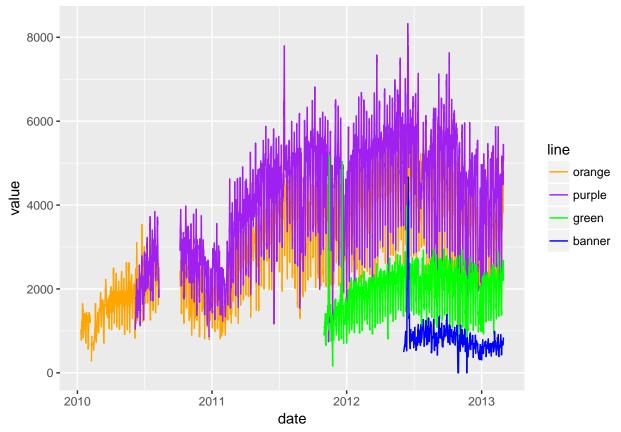
```
require(ggplot2)
### let's make a "ggplot"
### the format is ggplot(dataframe, aes(x=COLNAME, y=COLNAME))
### where COLNAME are colnames of the dataframe
### you can also set color to a different factor
### other options in AES (fill, alpha level -which is the "transparency" of points)
g <- ggplot(long.charm, 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()</pre>
```

## Warning: Removed 5328 rows containing missing values (geom\_point).



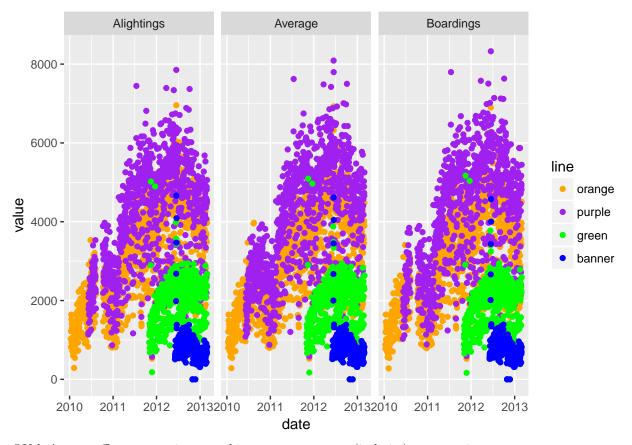
### Let's make Lines!
g + geom\_line()





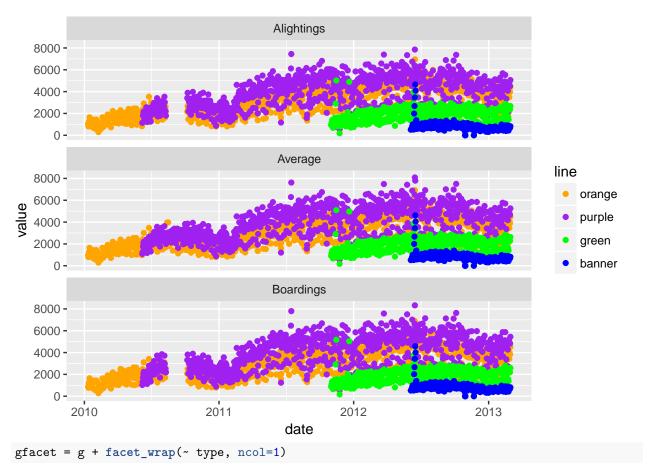
```
### let's make a new plot of poitns
gpoint <- g + geom_point()
### let's plot the value by the type of value - boardings/average, etc
gpoint + facet_wrap(~ type)</pre>
```

## Warning: Removed 5328 rows containing missing values (geom\_point).



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

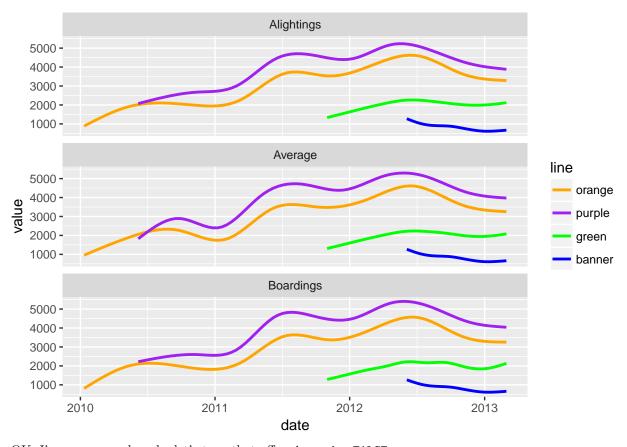
```
## let's compare vertically
gpoint + facet_wrap(~ type, ncol=1)
```



We can also smooth the data to give us a overall idea of how the average changes over time. I don't want to do a standard error (se).

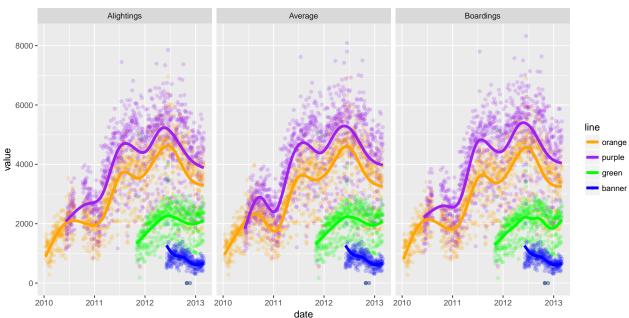
```
## let's smooth this - get a rough estimate of what's going on
gfacet + geom_smooth(se=FALSE)
```

## `geom\_smooth()` using method = 'gam'



OK, I've seen enough code, let's turn that off, using echo=FALSE.

## `geom\_smooth()` using method = 'gam'



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

