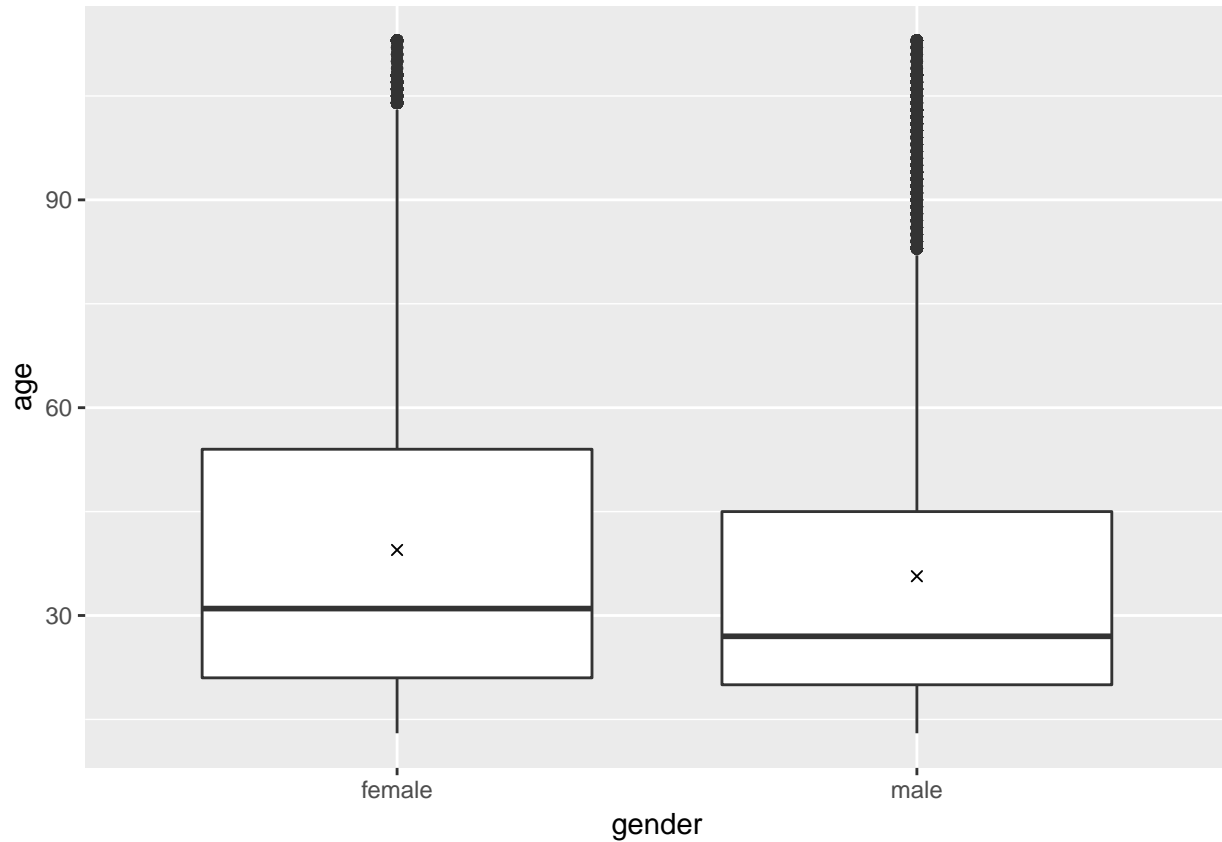


By: Reshu Singh

```
library(ggplot2)
pf <- read.csv('/home/reshu/Desktop/eda/lesson3/pseudo_facebook.tsv', sep = '\t')
```

```
ggplot(aes(x = gender, y = age),
       data = subset(pf, !is.na(gender))) + geom_boxplot() + stat_summary(fun.y = mean, geom = 'point',
```



```
ggplot(aes(x = age, y = friend_count),
       data = subset(pf, !is.na(gender))) + geom_line(aes(color = gender), stat = 'summary', fun.y = me
```



```
library(dplyr)
```

```
##
## Attaching package: 'dplyr'
##
## The following objects are masked from 'package:stats':
##
##   filter, lag
##
## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union
```

```
#chain functions together %>%
pf.fc_by_age_gender <- pf %>%
  filter(!is.na(gender)) %>%
  group_by(age, gender) %>%
  summarise(mean_friend_count = mean(friend_count),
            median_friend_count = median(friend_count),
            n = n()) %>%
  ungroup() %>%
  arrange(age)
```

```
head(pf.fc_by_age_gender)
```

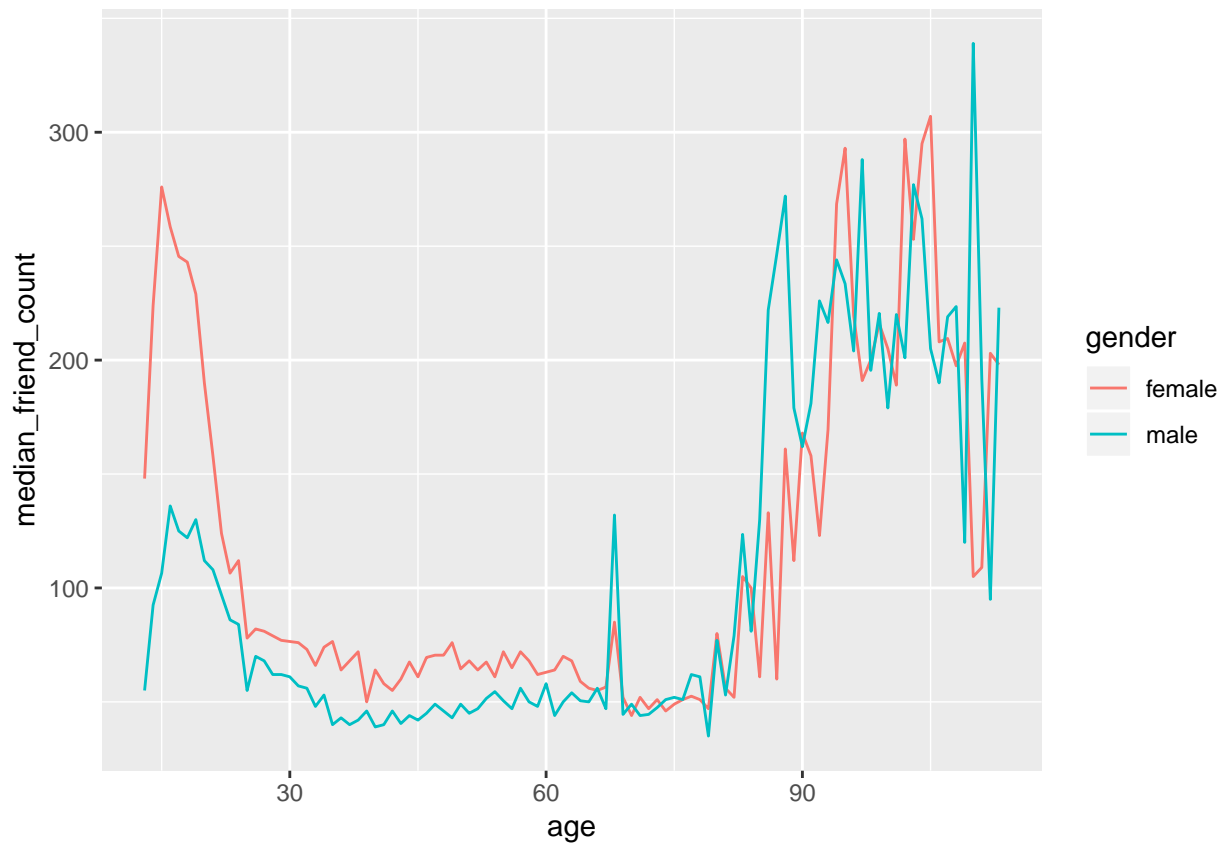
```
## # A tibble: 6 x 5
##   age gender mean_friend_count median_friend_count     n
##   <int> <fct>         <dbl>           <dbl> <int>
## 1    13 female           259.             148    193
```

## 2	13 male	102.	55	291
## 3	14 female	362.	224	847
## 4	14 male	164.	92.5	1078
## 5	15 female	539.	276	1139
## 6	15 male	201.	106.	1478

Plotting Conditional Summaries

Notes:

```
ggplot(aes(x = age, y = median_friend_count),
  data = pf.fc_by_age_gender) +
  geom_line(aes(color = gender))
```



Wide and Long Format

Notes:

Notes:

```
install.packages("tidyr")
```

```
## Installing package into '/home/reshu/R/x86_64-pc-linux-gnu-library/3.4'
## (as 'lib' is unspecified)
```

```
library(tidyr)

spread(subset(pf.fc_by_age_gender,
  select = c('gender', 'age', 'median_friend_count')),
  gender, median_friend_count)
```

```
## # A tibble: 101 x 3
##   age female male
##   <int> <dbl> <dbl>
## 1    13   148   55
## 2    14   224  92.5
## 3    15   276 106.
## 4    16   258. 136
## 5    17   246. 125
## 6    18   243  122
## 7    19   229  130
## 8    20   190  112
## 9    21   158  108
## 10   22   124   97
## # ... with 91 more rows
```

Reshaping Data

Notes:

```
install.packages('reshape2')
```

```
## Installing package into '/home/reshu/R/x86_64-pc-linux-gnu-library/3.4'
## (as 'lib' is unspecified)
```

```
library(reshape2)
```

```
##
## Attaching package: 'reshape2'
## The following object is masked from 'package:tidyr':
##
## smiths
```

```
pf.fc_by_age_gender.wide <-
  subset(pf.fc_by_age_gender[c('age', 'gender', 'median_friend_count')],
    !is.na(gender)) %>%
  spread(gender, median_friend_count) %>%
  mutate(ratio = male / female)
```

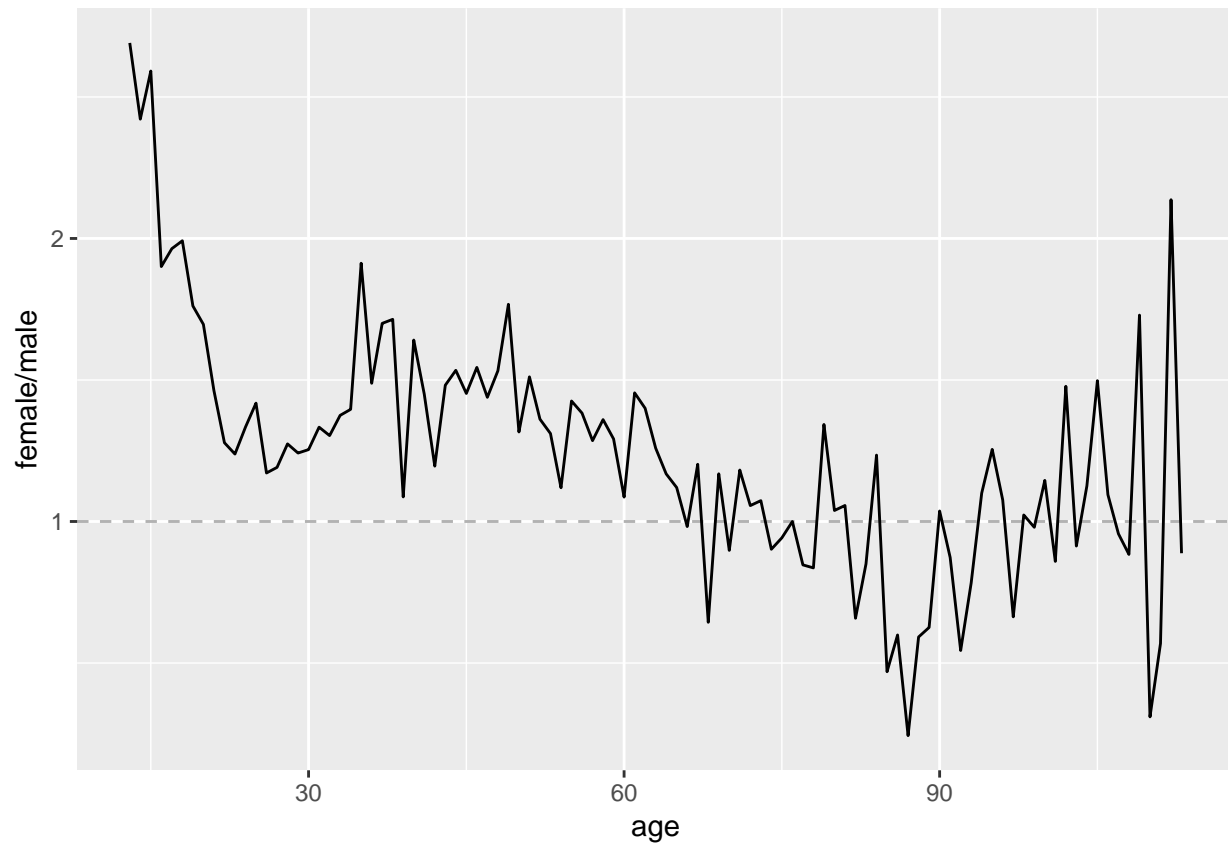
```
head(pf.fc_by_age_gender.wide)
```

```
## # A tibble: 6 x 4
##   age female male ratio
##   <int> <dbl> <dbl> <dbl>
## 1    13   148   55 0.372
## 2    14   224  92.5 0.413
## 3    15   276 106. 0.386
## 4    16   258. 136 0.526
## 5    17   246. 125 0.509
## 6    18   243  122 0.502
```

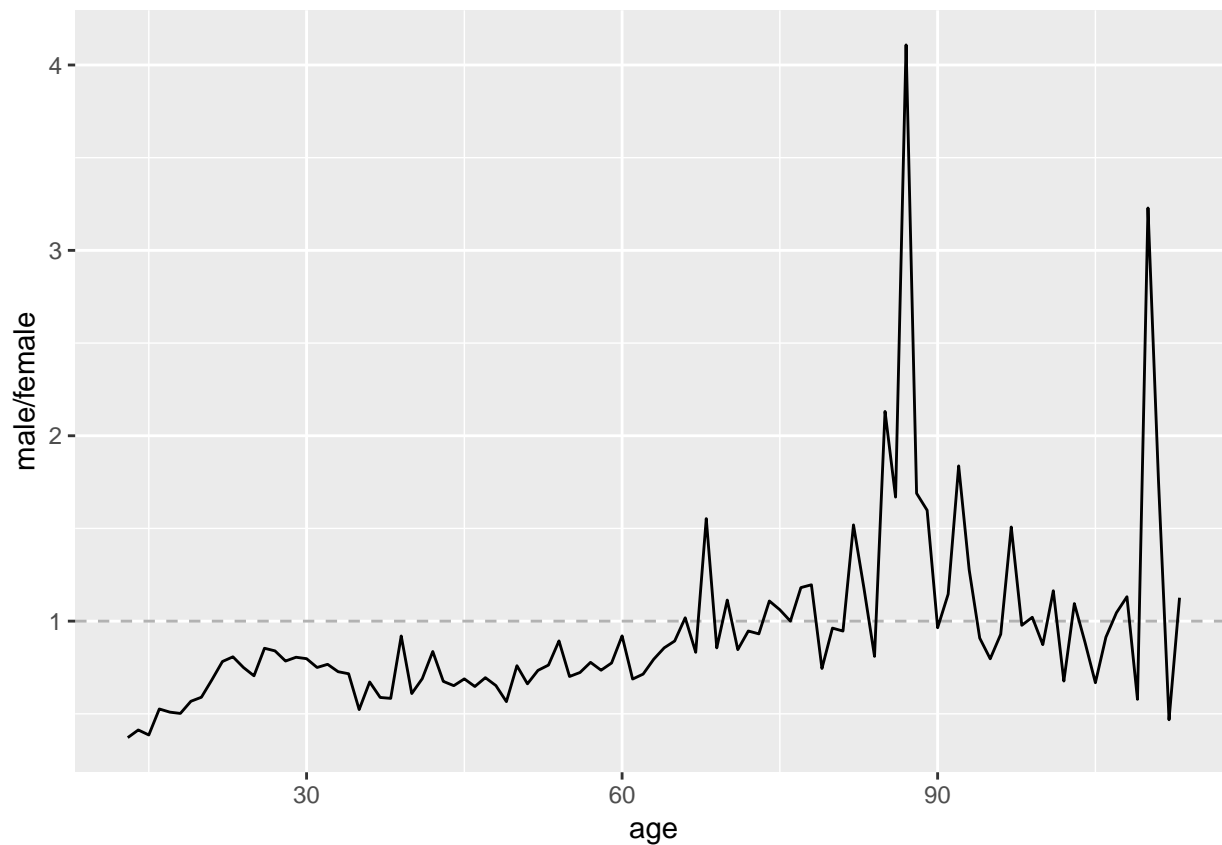
Ratio Plot

Notes:

```
ggplot(aes(x = age, y = female / male),  
       data = pf.fc_by_age_gender.wide) +  
  geom_line() +  
  geom_hline(yintercept = 1, alpha = 0.3, linetype = 2)
```



```
ggplot(aes(x = age, y = male / female),  
       data = pf.fc_by_age_gender.wide) +  
  geom_line() +  
  geom_hline(yintercept = 1, alpha = 0.3, linetype = 2)
```



Third Quantitative Variable

Notes:

```
pf$year_joined <- floor(2014 - pf$tenure/365)
```

Cut a Variable

Notes:

```
summary(pf$year_joined)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.   NA's
##      2005    2012    2012    2012    2013    2014     2
```

```
table(pf$year_joined)
```

```
##
##  2005  2006  2007  2008  2009  2010  2011  2012  2013  2014
##    9    15   581  1507  4557  5448  9860 33366 43588   70
```

```
pf$year_joined.bucket <- cut(pf$year_joined, c(2004, 2009, 2011, 2012, 2014))
```

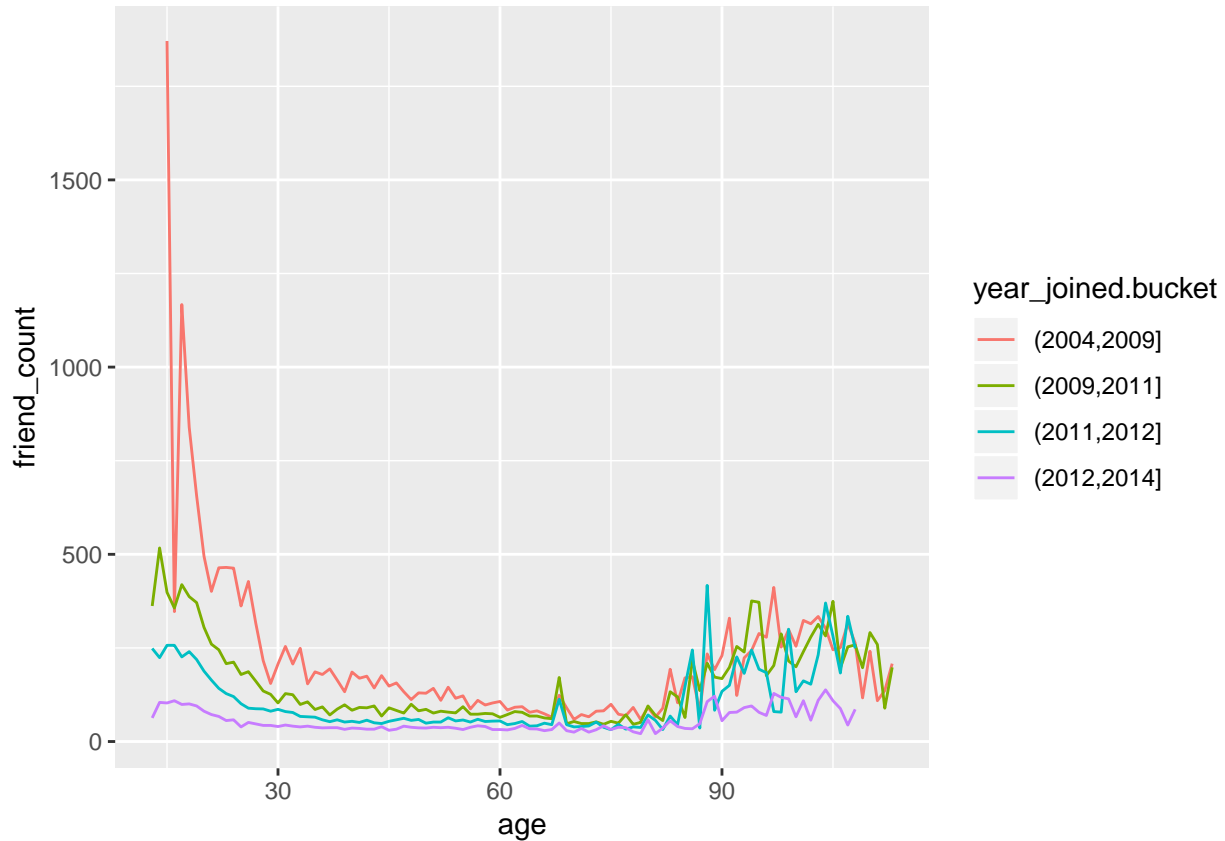
Plotting it All Together

Notes:

```
table(pf$year_joined.bucket, useNA = 'ifany')
```

```
##  
## (2004,2009] (2009,2011] (2011,2012] (2012,2014] <NA>  
##      6669      15308      33366      43658      2
```

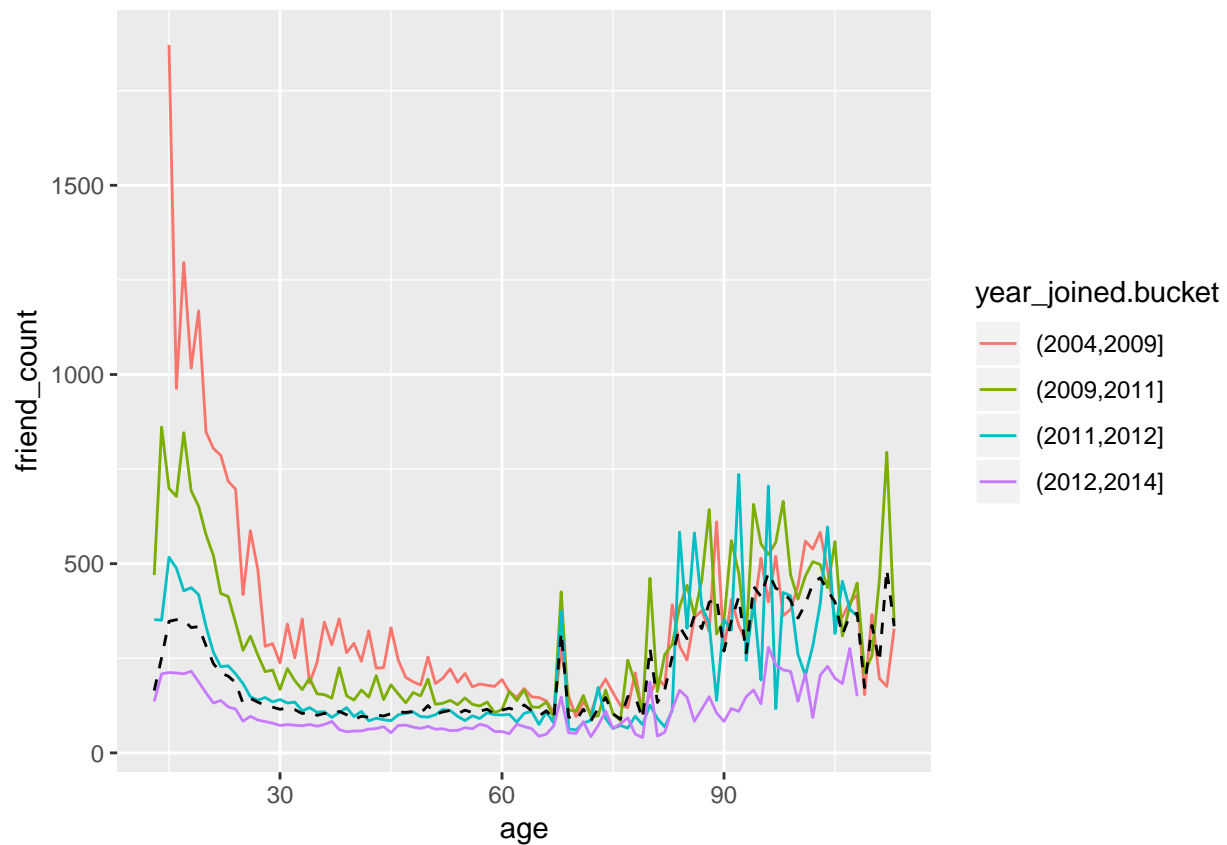
```
ggplot(aes(x = age, y = friend_count),  
       data = subset(pf, !is.na(year_joined.bucket))) + geom_line(aes(color = year_joined.bucket), stat = 'summary')
```



Plot the Grand Mean

Notes:

```
ggplot(aes(x = age, y = friend_count),  
       data = subset(pf, !is.na(year_joined.bucket))) + geom_line(aes(color = year_joined.bucket), stat = 'summary',  
       geom_line(stat = 'summary', fun.y = mean, linetype = 2))
```



Friending Rate

Notes:

```
with(subset(pf, tenure >= 1), summary(friend_count / tenure))
```

##	Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
##	0.0000	0.0775	0.2205	0.6096	0.5658	417.0000

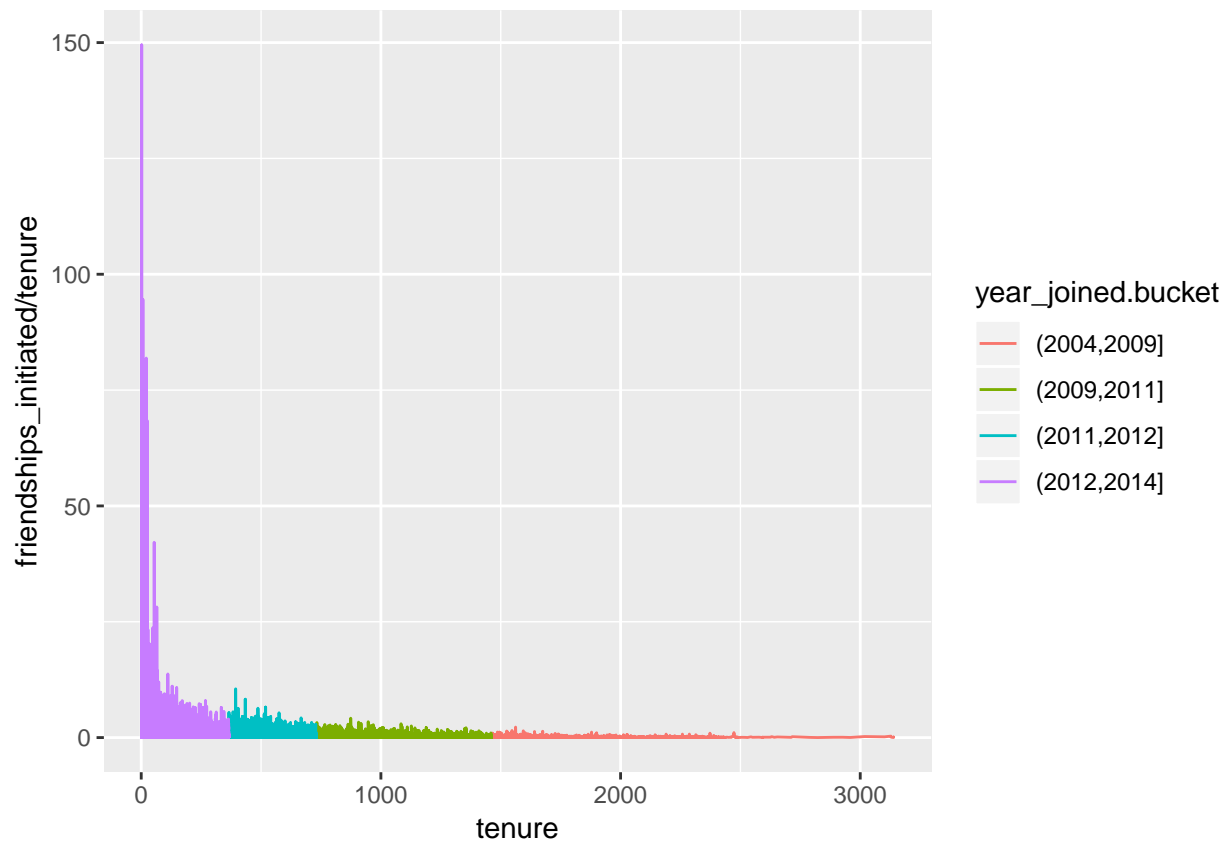
Friendships Initiated

Notes:

What is the median friend rate? 0.2205

What is the maximum friend rate? 417

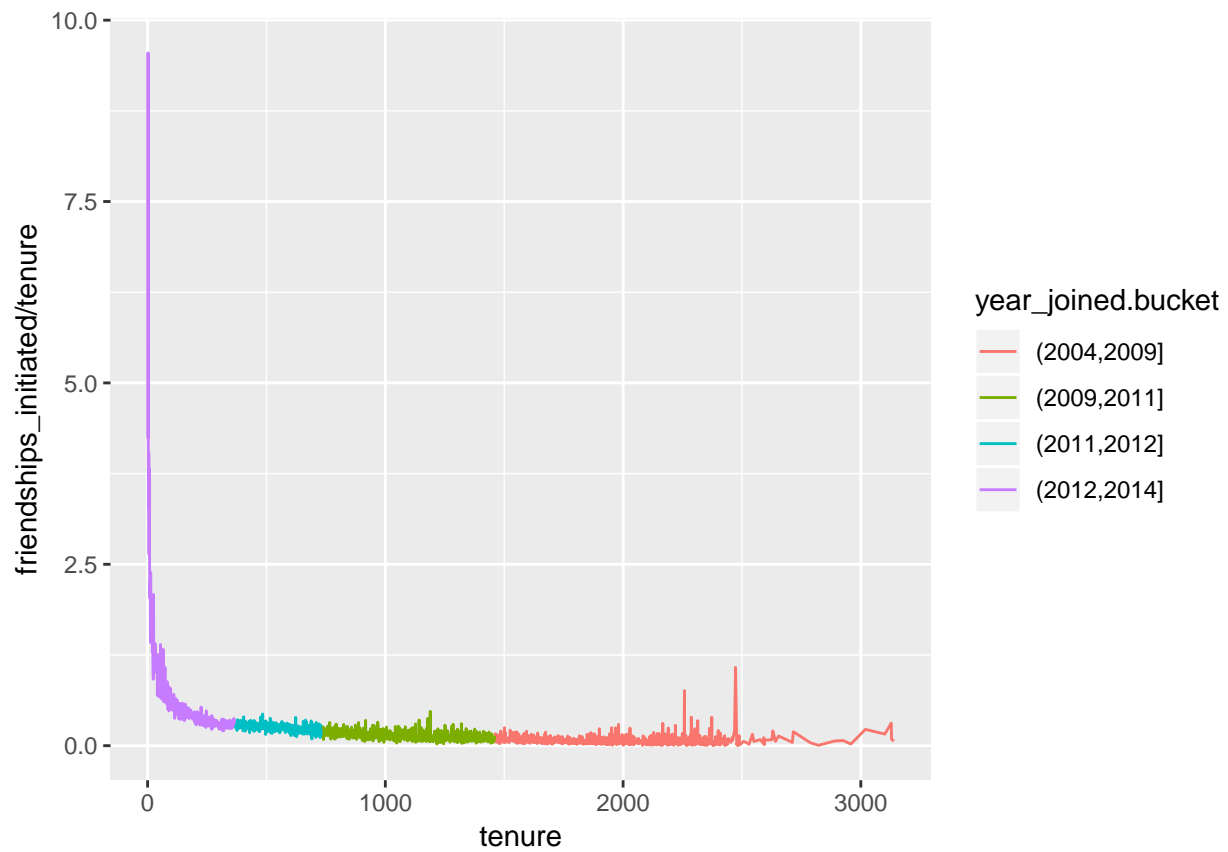
```
ggplot(aes(x = tenure, y = friendships_initiated / tenure),
       data = subset(pf, tenure >= 1)) +
  geom_line(aes(color = year_joined.bucket))
```

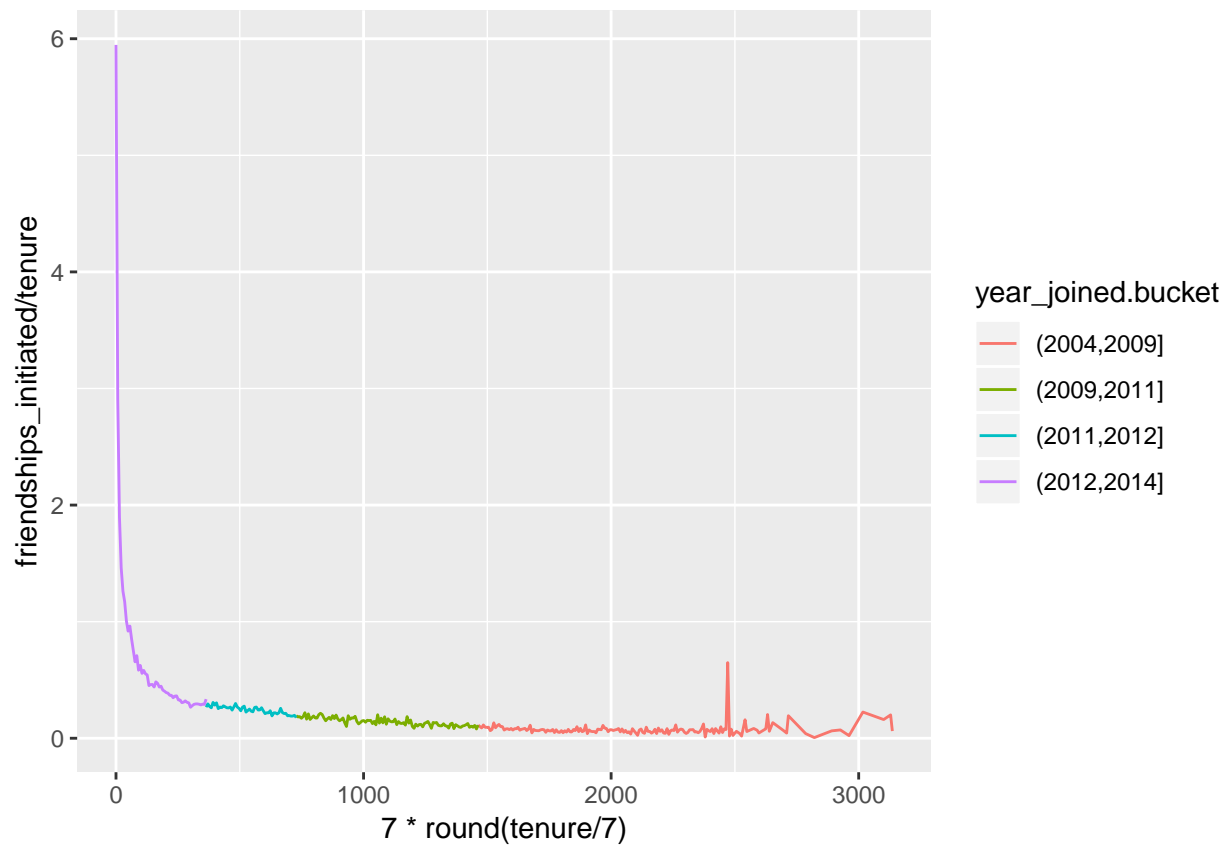
Bias-Variance Tradeoff Revisited

Notes:

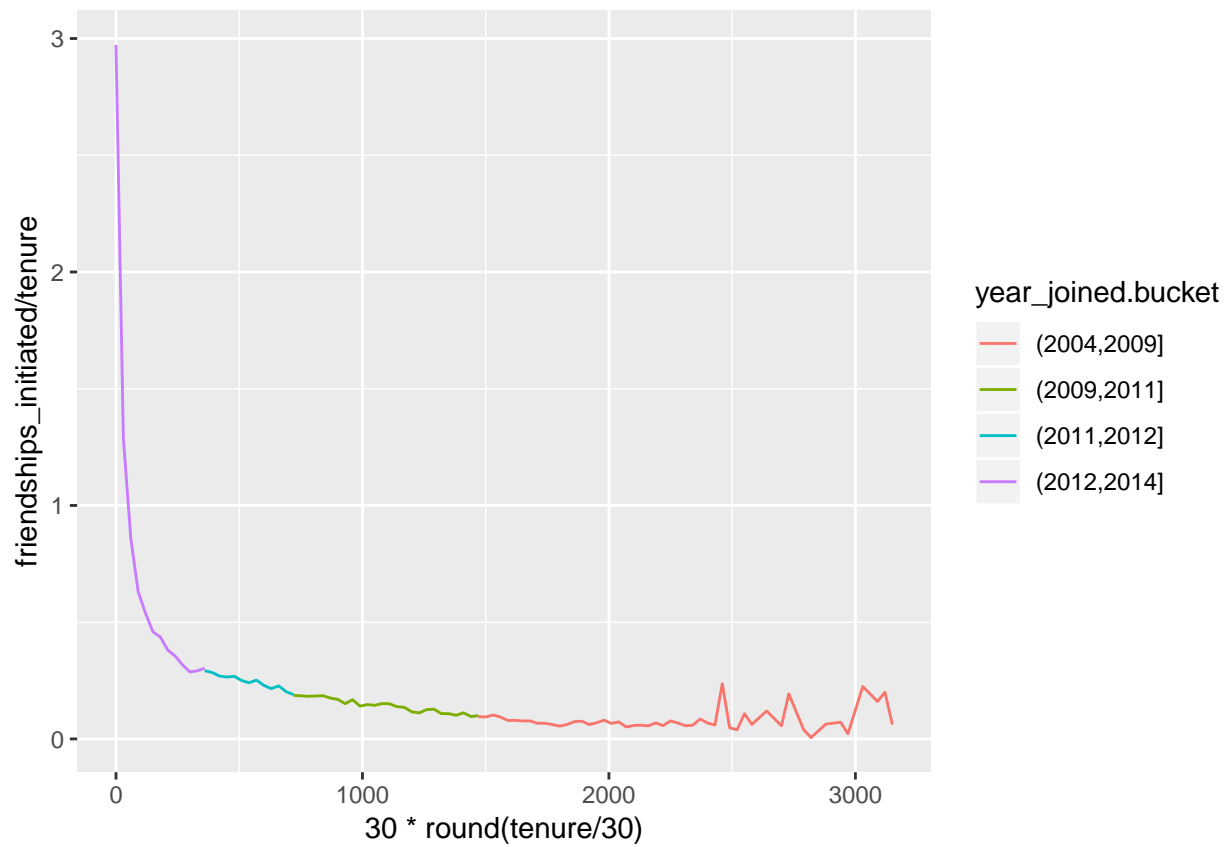
```
ggplot(aes(x = tenure, y = friendships_initiated / tenure),  
  data = subset(pf, tenure >= 1)) +  
  geom_line(aes(color = year_joined.bucket),  
    stat = 'summary',  
    fun.y = mean)
```



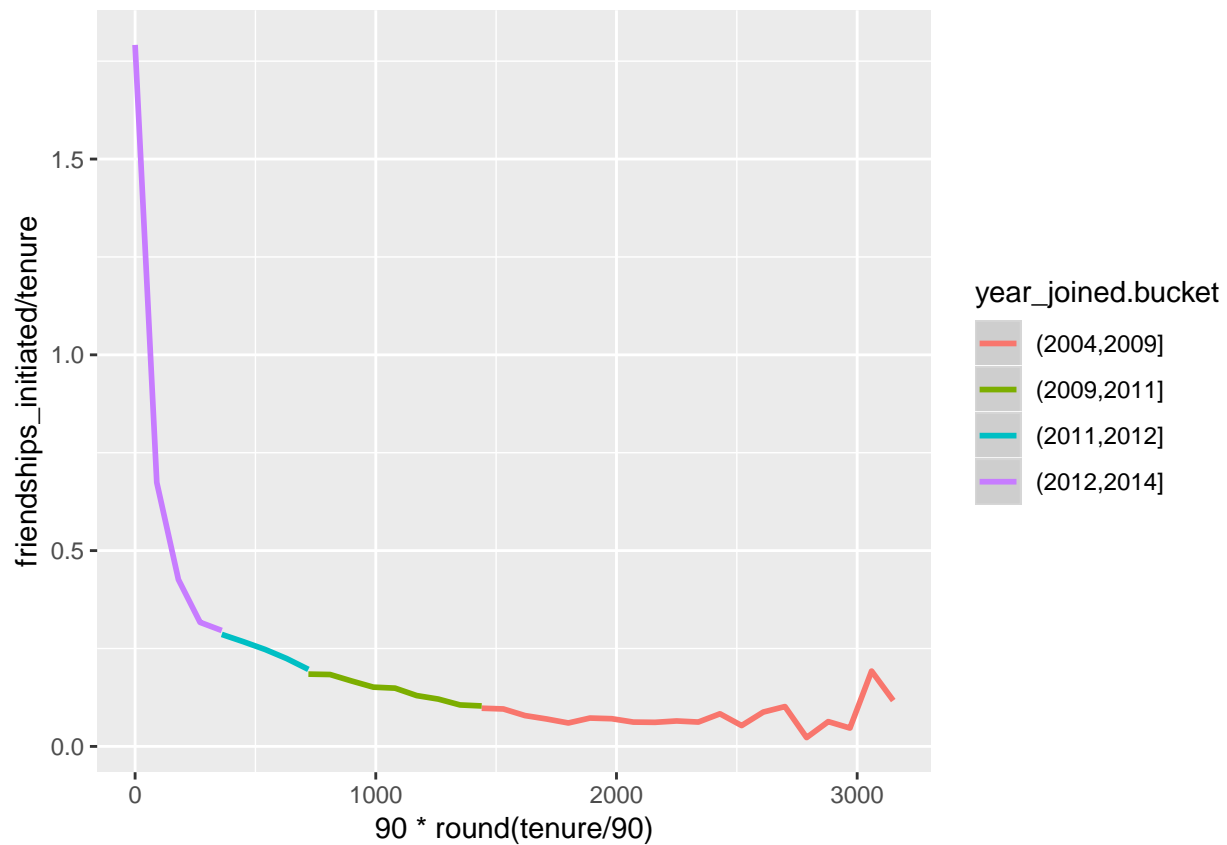
```
ggplot(aes(x = 7 * round(tenure / 7), y = friendships_initiated / tenure),
  data = subset(pf, tenure > 0)) +
  geom_line(aes(color = year_joined.bucket),
    stat = "summary",
    fun.y = mean)
```



```
ggplot(aes(x = 30 * round(tenure / 30), y = friendships_initiated / tenure),
  data = subset(pf, tenure > 0)) +
  geom_line(aes(color = year_joined.bucket),
    stat = "summary",
    fun.y = mean)
```



```
ggplot(aes(x = 90 * round(tenure / 90), y = friendships_initiated / tenure),
  data = subset(pf, tenure > 0)) +
  geom_smooth(aes(color = year_joined.bucket),
    stat = "summary",
    fun.y = mean)
```



Sean's NFL Fan Sentiment Study

Notes:

Introducing the Yogurt Data Set

Notes:

Histograms Revisited

Notes:

```
yo <- read.csv('/home/reshu/Desktop/eda/lesson5/yogurt.csv')
str(yo)
```

```
## 'data.frame': 2380 obs. of 9 variables:
## $ obs : int 1 2 3 4 5 6 7 8 9 10 ...
## $ id : int 2100081 2100081 2100081 2100081 2100081 2100081 2100081 2100081 2100081 2100081 ...
## $ time : int 9678 9697 9825 9999 10015 10029 10036 10042 10083 10091 ...
## $ strawberry : int 0 0 0 0 1 1 0 0 0 0 ...
## $ blueberry : int 0 0 0 0 0 0 0 0 0 0 ...
```

```
## $ pina.colada: int  0 0 0 0 1 2 0 0 0 0 ...
## $ plain       : int  0 0 0 0 0 0 0 0 0 0 ...
## $ mixed.berry: int  1 1 1 1 1 1 1 1 1 1 ...
## $ price       : num  59 59 65 65 49 ...
```

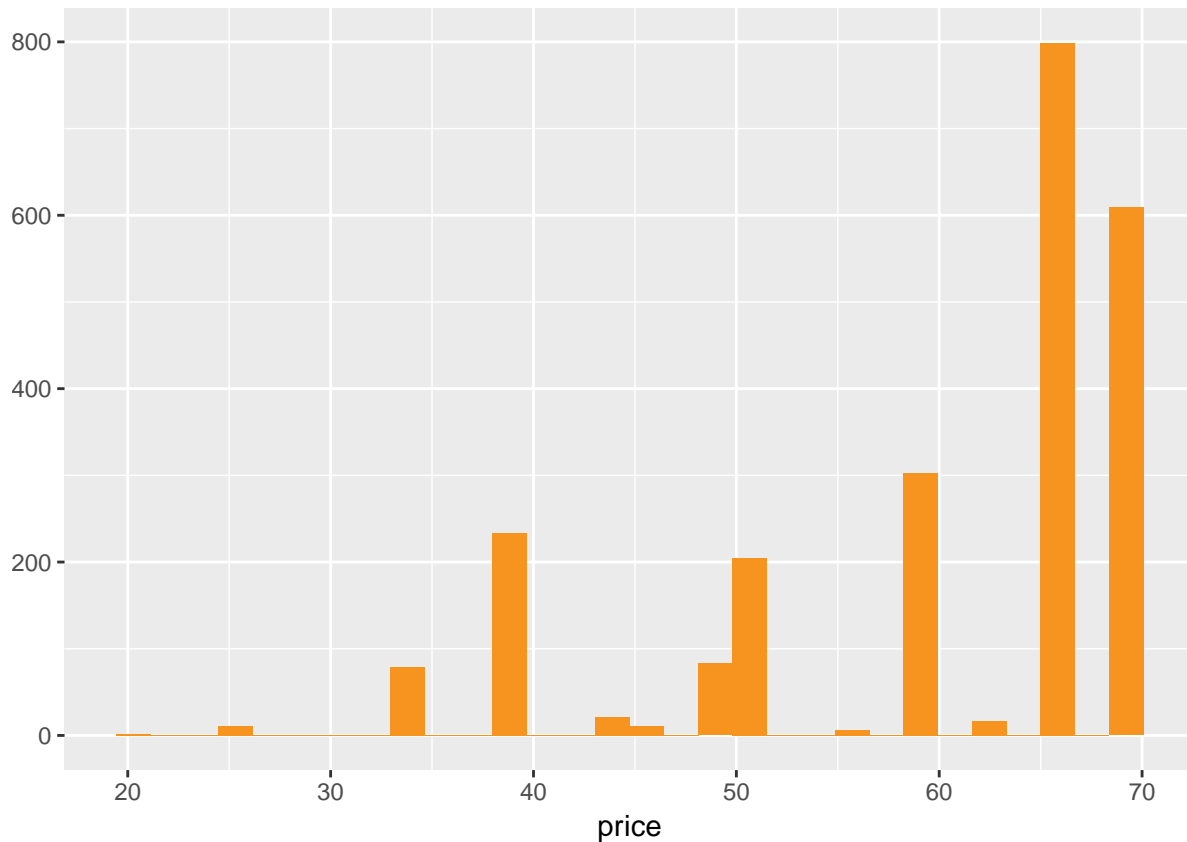
```
#Change the id from an int to a factor
```

```
yo$id <- factor(yo$id)
str(yo)
```

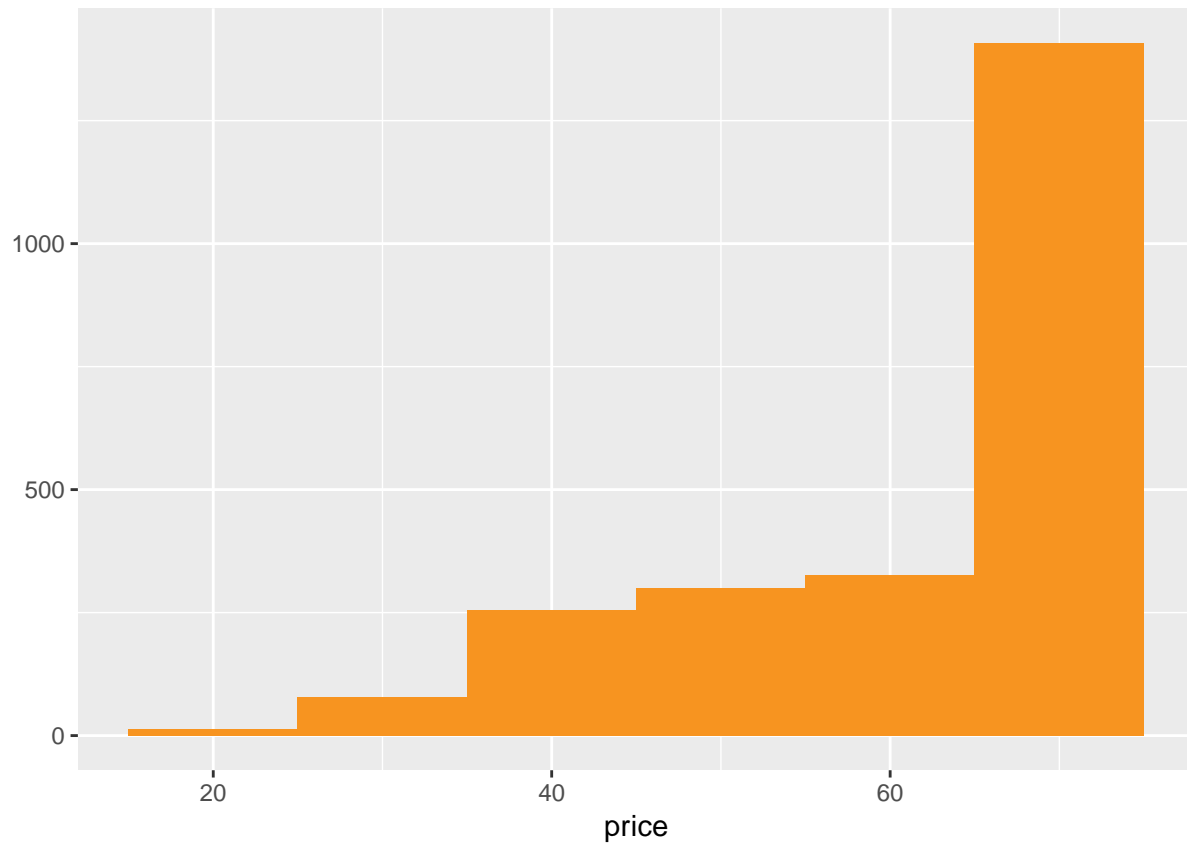
```
## 'data.frame':  2380 obs. of  9 variables:
## $ obs       : int  1 2 3 4 5 6 7 8 9 10 ...
## $ id        : Factor w/ 332 levels "2100081","2100370",...: 1 1 1 1 1 1 1 1 1 1 ...
## $ time      : int  9678 9697 9825 9999 10015 10029 10036 10042 10083 10091 ...
## $ strawberry: int  0 0 0 0 1 1 0 0 0 0 ...
## $ blueberry : int  0 0 0 0 0 0 0 0 0 0 ...
## $ pina.colada: int  0 0 0 0 1 2 0 0 0 0 ...
## $ plain     : int  0 0 0 0 0 0 0 0 0 0 ...
## $ mixed.berry: int  1 1 1 1 1 1 1 1 1 1 ...
## $ price     : num  59 59 65 65 49 ...
```

```
qplot(data = yo, x = price, fill = I('#F79420'))
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```



```
qplot(data = yo, x = price, fill = I('#F79420'), binwidth = 10)
```



Number of Purchases

Notes:

`summary(yo)`

```
##      obs      id      time      strawberry
## Min.   : 1.0 2132290: 74 Min.   : 9662 Min.   : 0.0000
## 1st Qu.: 696.5 2130583: 59 1st Qu.: 9843 1st Qu.: 0.0000
## Median :1369.5 2124073: 50 Median :10045 Median : 0.0000
## Mean   :1367.8 2149500: 50 Mean   :10050 Mean   : 0.6492
## 3rd Qu.:2044.2 2101790: 47 3rd Qu.:10255 3rd Qu.: 1.0000
## Max.   :2743.0 2129528: 39 Max.   :10459 Max.   :11.0000
##      (Other):2061
##      blueberry      pina.colada      plain      mixed.berry
## Min.   : 0.0000 Min.   : 0.0000 Min.   :0.0000 Min.   :0.0000
## 1st Qu.: 0.0000 1st Qu.: 0.0000 1st Qu.:0.0000 1st Qu.:0.0000
## Median : 0.0000 Median : 0.0000 Median :0.0000 Median :0.0000
## Mean   : 0.3571 Mean   : 0.3584 Mean   :0.2176 Mean   :0.3887
## 3rd Qu.: 0.0000 3rd Qu.: 0.0000 3rd Qu.:0.0000 3rd Qu.:0.0000
## Max.   :12.0000 Max.   :10.0000 Max.   :6.0000 Max.   :8.0000
##
##      price
## Min.   :20.00
## 1st Qu.:50.00
```

```
## Median :65.04
## Mean   :59.25
## 3rd Qu.:68.96
## Max.   :68.96
##
```

```
length(unique(yo$price))
```

```
## [1] 20
```

```
table(yo$price)
```

```
##
##      20 24.96 33.04  33.2 33.28 33.36 33.52 39.04      44 45.04 48.96 49.52
##      2   11   54    1    1   22    1   234    21   11   81    1
## 49.6   50 55.04 58.96   62 63.04 65.04 68.96
##      1   205    6   303   15    2   799   609
```

```
str(yo)
```

```
## 'data.frame': 2380 obs. of 9 variables:
## $ obs      : int  1 2 3 4 5 6 7 8 9 10 ...
## $ id       : Factor w/ 332 levels "2100081","2100370",...: 1 1 1 1 1 1 1 1 1 1 ...
## $ time     : int  9678 9697 9825 9999 10015 10029 10036 10042 10083 10091 ...
## $ strawberry : int  0 0 0 0 1 1 0 0 0 0 ...
## $ blueberry  : int  0 0 0 0 0 0 0 0 0 0 ...
## $ pina.colada: int  0 0 0 0 1 2 0 0 0 0 ...
## $ plain      : int  0 0 0 0 0 0 0 0 0 0 ...
## $ mixed.berry: int  1 1 1 1 1 1 1 1 1 1 ...
## $ price      : num  59 59 65 65 49 ...
```

```
yo$all.purchases
```

```
## NULL
```

```
yo <- transform(yo, all.purchases = strawberry + blueberry + pina.colada + plain + mixed.berry)
```

```
summary(yo$all.purchases)
```

```
##      Min. 1st Qu.  Median      Mean 3rd Qu.      Max.
##      1.000  1.000   2.000   1.971   2.000   21.000
```

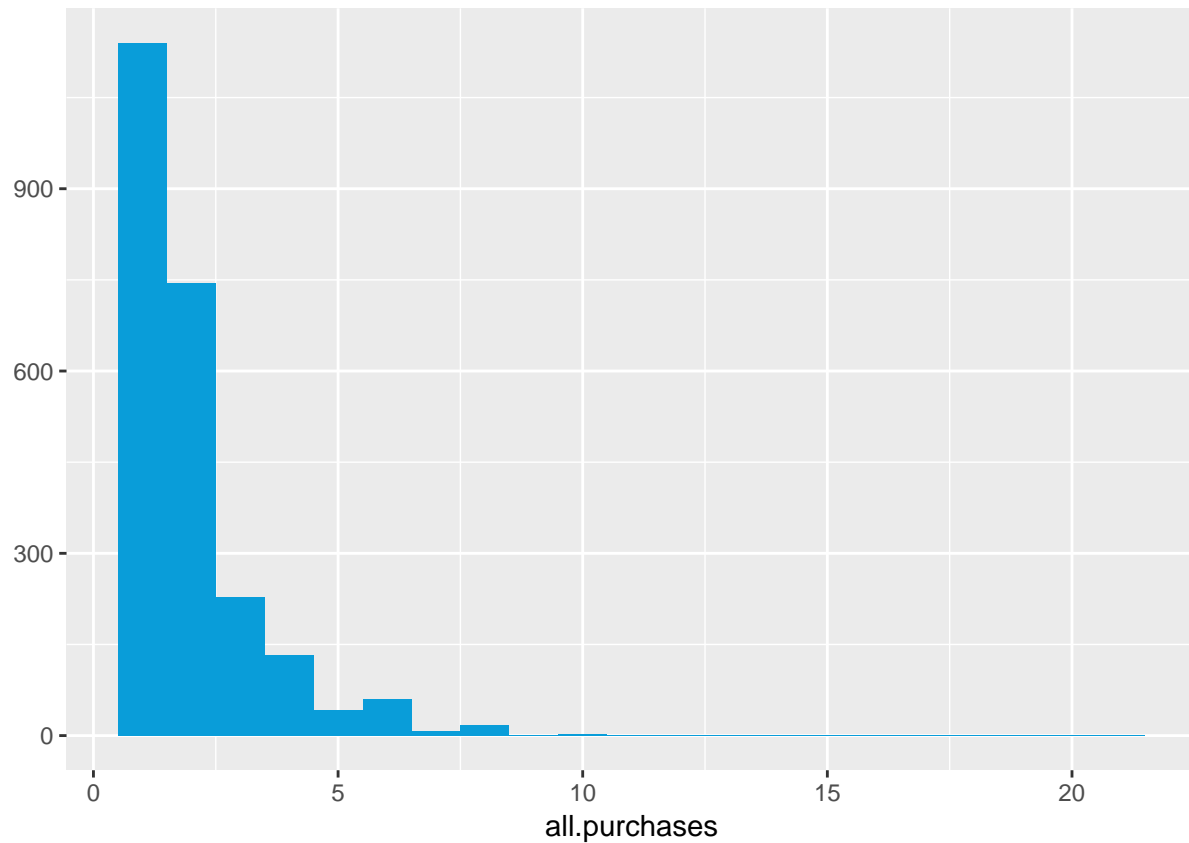
```
#Alternate way
```

```
yo$all.purchases <- yo$strawberry + yo$blueberry + yo$pina.colada + yo$plain + yo$mixed.berry
```

Prices over Time

Notes:

```
qplot(x = all.purchases, data = yo, binwidth = 1, fill = I('#099dd9'))
```

Sampling Observations

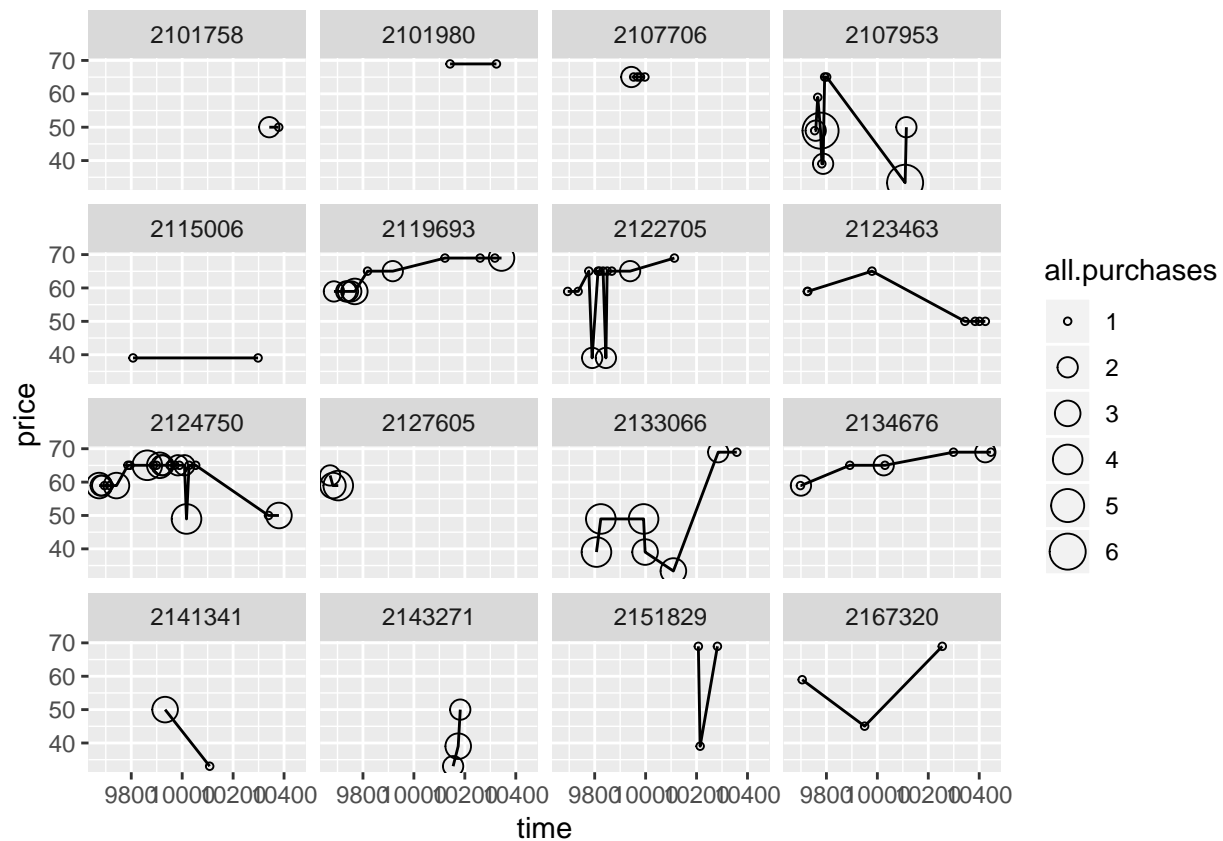
Notes:

Looking at Samples of Households

```
#Set the seed for reproducible results

set.seed(4230)
sample.ids <- sample(levels(yo$id), 16)

ggplot(aes(x = time, y = price),
  data = subset(yo, id %in% sample.ids)) +
  facet_wrap( ~ id) +
  geom_line() +
  geom_point(aes(size = all.purchases), pch = 1)
```



The Limits of Cross Sectional Data

Notes:

Many Variables

Notes:

Even More Variables

Notes:

Heat Maps

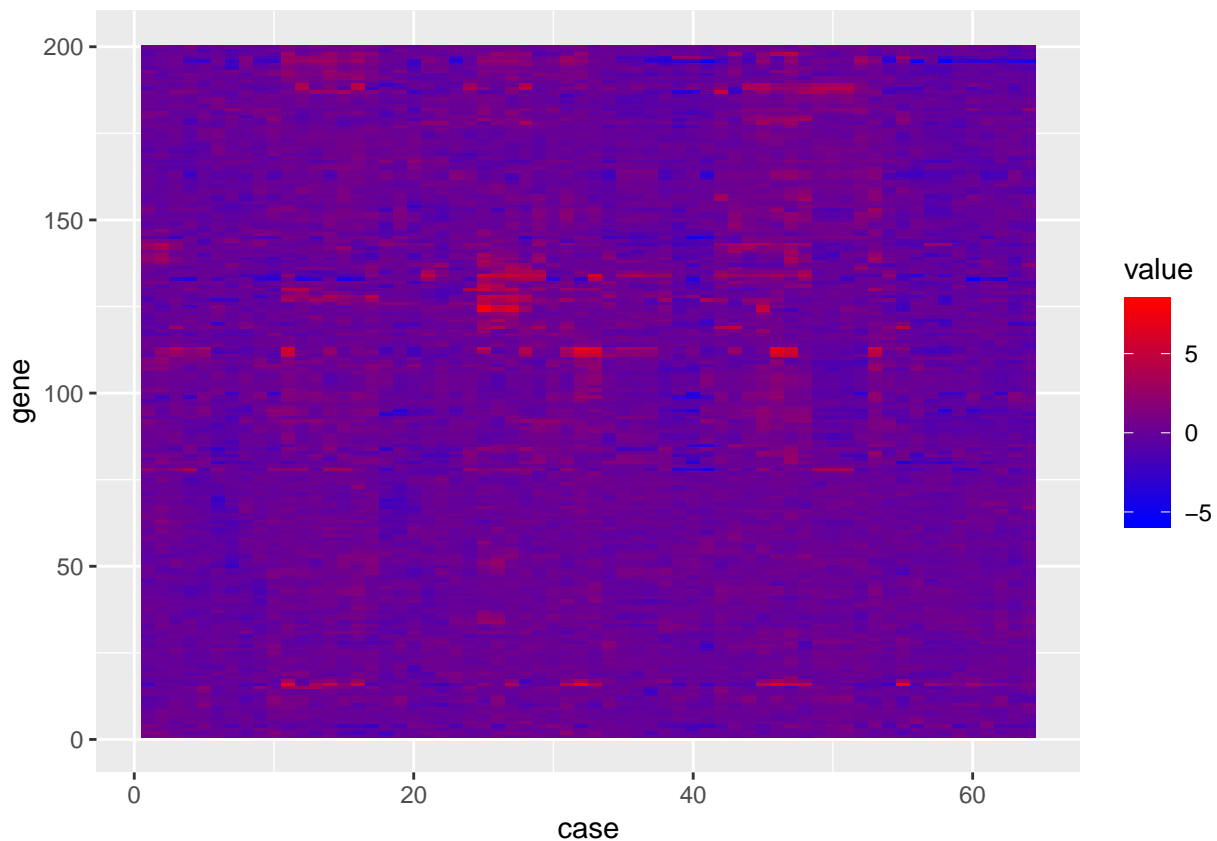
Notes:

```
nci <- read.table("/home/reshu/Desktop/eda/lesson5/nci.tsv")
colnames(nci) <- c(1:64)
```

```
nci.long.samp <- melt(as.matrix(nci[1:200,]))
names(nci.long.samp) <- c("gene", "case", "value")
head(nci.long.samp)
```

```
##   gene case  value
## 1    1    1  0.300
## 2    2    1  1.180
## 3    3    1  0.550
## 4    4    1  1.140
## 5    5    1 -0.265
## 6    6    1 -0.070
```

```
ggplot(aes(y = gene, x = case, fill = value),
  data = nci.long.samp) +
  geom_tile() +
  scale_fill_gradientn(colours = colorRampPalette(c("blue", "red"))(100))
```



Analyzing Three or More Variables

Reflection:

Click **KnitHTML** to see all of your hard work and to have an html page of this lesson, your answers, and your notes!