Homework 2

The data set calif_penn_2011.csv contains information about the housing stock of California and Pennsylvania, as of 2011. Information as aggregated into "Census tracts", geographic regions of a few thousand people which are supposed to be fairly homogeneous economically and socially.

1. *Loading and cleaning*
    a. Load the data into a dataframe called `ca_pa`.

```
setwd("D:/github/Rlab/")
ca_pa<-read.csv("data/calif_penn_2011.csv",header=T)
ca_pa = as.data.frame(ca_pa)
```

b. How many rows and columns does the dataframe have?

```
dim(ca_pa)#11275 rows and 34 cloumns
```

```
## [1] 11275    34
```

c. Run this command, and explain, in words, what this does:

```
colSums(apply(ca_pa,c(1,2),is.na))
```

```
##                           X                      GEO.id2
##                           0                            0
##                     STATEFP                     COUNTYFP
##                           0                            0
##                     TRACTCE                   POPULATION
##                           0                            0
##                    LATITUDE                    LONGITUDE
##                           0                            0
##           GEO.display.label          Median_house_value
##                           0                          599
##                Total_units                Vacant_units
##                           0                            0
##                Median_rooms  Mean_household_size_owners
##                         157                          215
## Mean_household_size_renters         Built_2005_or_later
##                         152                           98
##          Built_2000_to_2004                 Built_1990s
##                          98                           98
##                 Built_1980s                 Built_1970s
##                          98                           98
##                 Built_1960s                 Built_1950s
##                          98                           98
##                 Built_1940s       Built_1939_or_earlier
##                          98                           98
##                  Bedrooms_0                  Bedrooms_1
##                          98                           98
##                  Bedrooms_2                  Bedrooms_3
##                          98                           98
##                  Bedrooms_4          Bedrooms_5_or_more
```

```
##                           98                           98
##                       Owners                      Renters
##                          100                          100
##     Median_household_income       Mean_household_income
##                          115                          126
```

find how many na in each column d. The function `na.omit()` takes a dataframe and returns a new dataframe, omitting any row containing an NA value. Use it to purge the data set of rows with incomplete data.

```
ca_pa = na.omit(ca_pa)
```

`e. How many rows did this eliminate?`

```
11275-dim(ca_pa)[1]
```

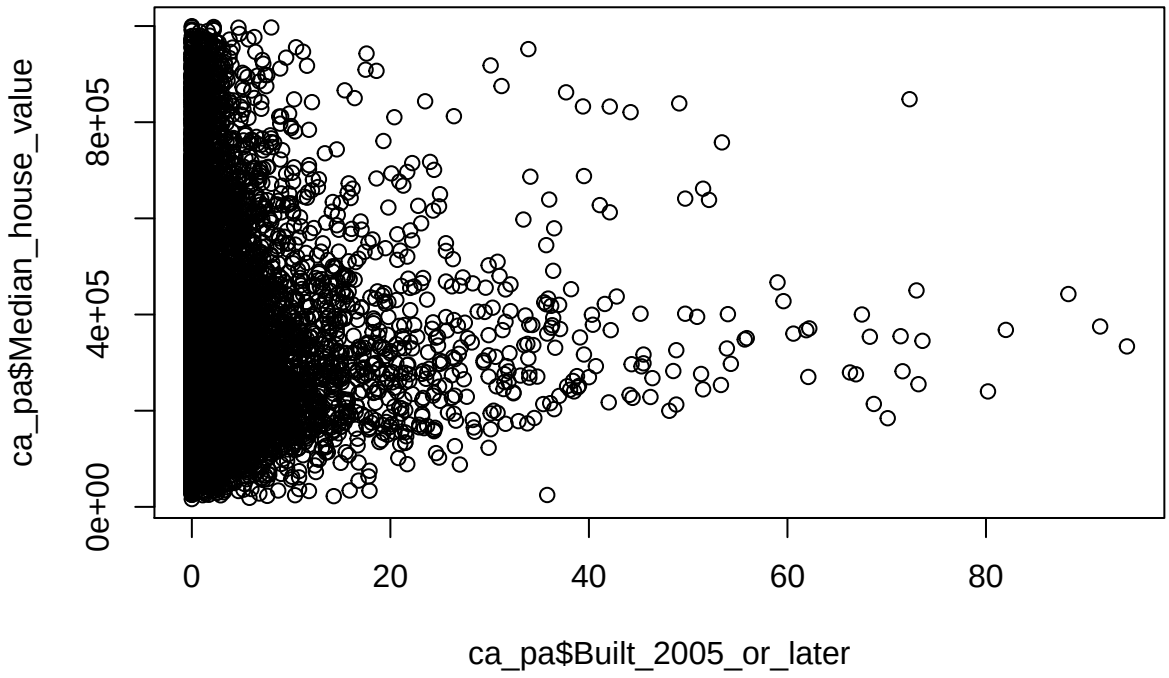
```
## [1] 670
```

`f. Are your answers in (c) and (e) compatible? Explain.`

Yes, since some rows have more than 1 NA, thus the number of NAs could be bigger than the number of rows that have NAs.

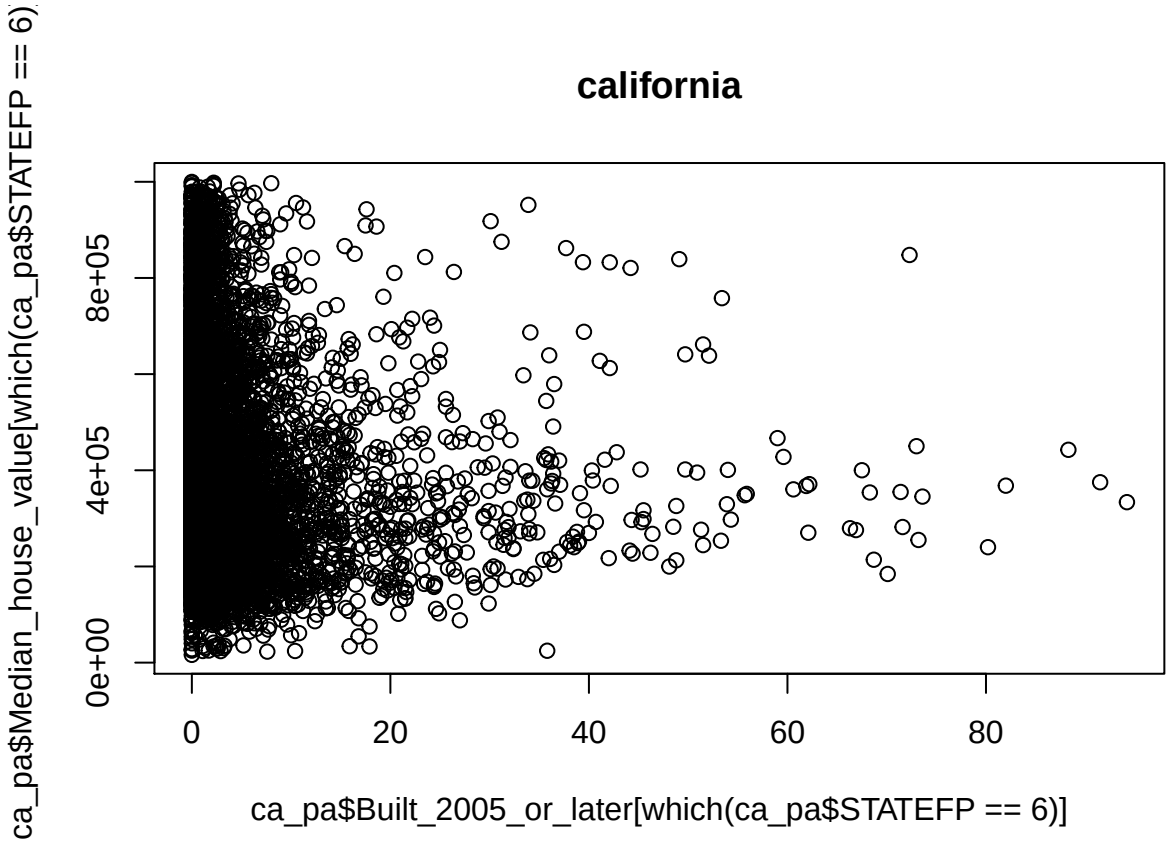
2. *This Very New House*
   a. The variable `Built_2005_or_later` indicates the percentage of houses in each Census tract built since 2005. Plot median house prices against this variable.

```
plot(x=ca_pa$Built_2005_or_later,y=ca_pa$Median_house_value)
```

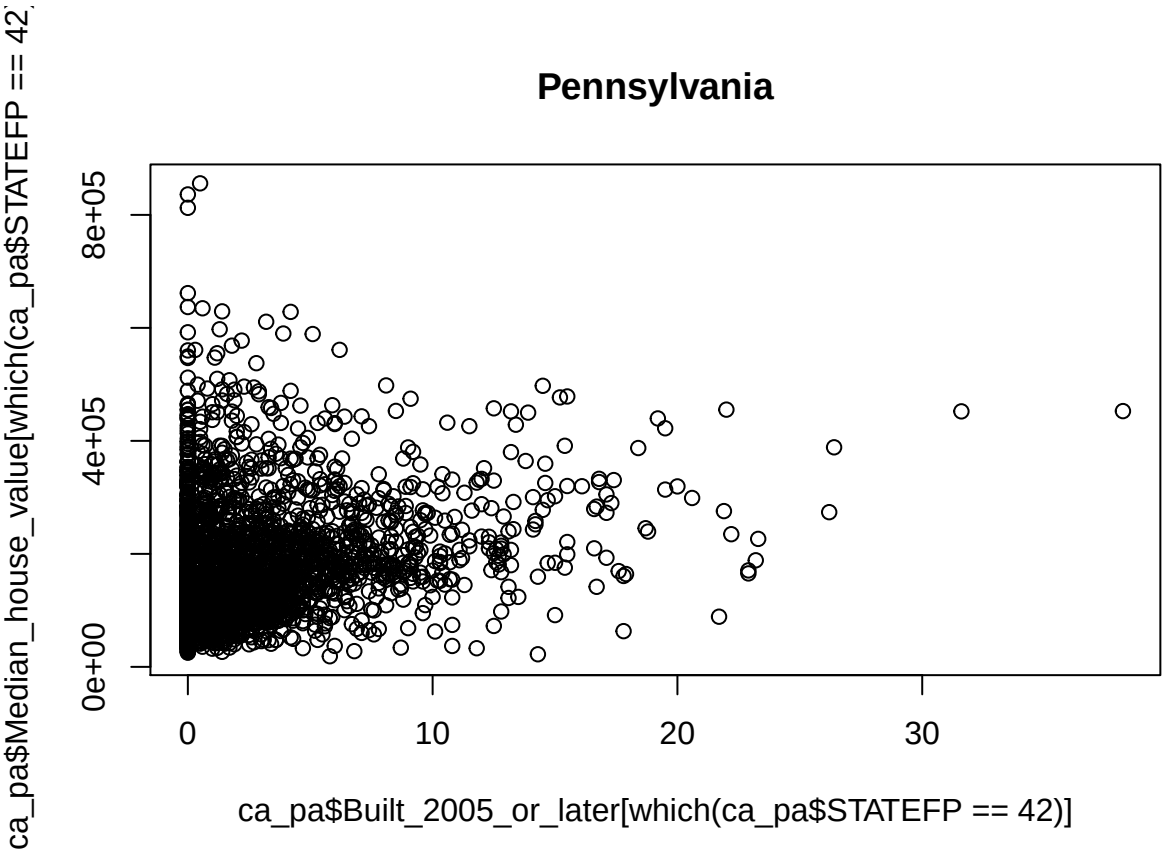


b. Make a new plot, or pair of plots, which breaks this out by state. Note that the state is recorded in the `STATEFP` variable, with California being state 6 and Pennsylvania state 42.

```
plot(x=ca_pa$Built_2005_or_later[which(ca_pa$STATEFP==6)],
     y=ca_pa$Median_house_value[which(ca_pa$STATEFP==6)])
title("california")
```




```
plot(x=ca_pa$Built_2005_or_later[which(ca_pa$STATEFP==42)],
     y=ca_pa$Median_house_value[which(ca_pa$STATEFP==42)])
title("Pennsylvania")
```



3. *Nobody Home*

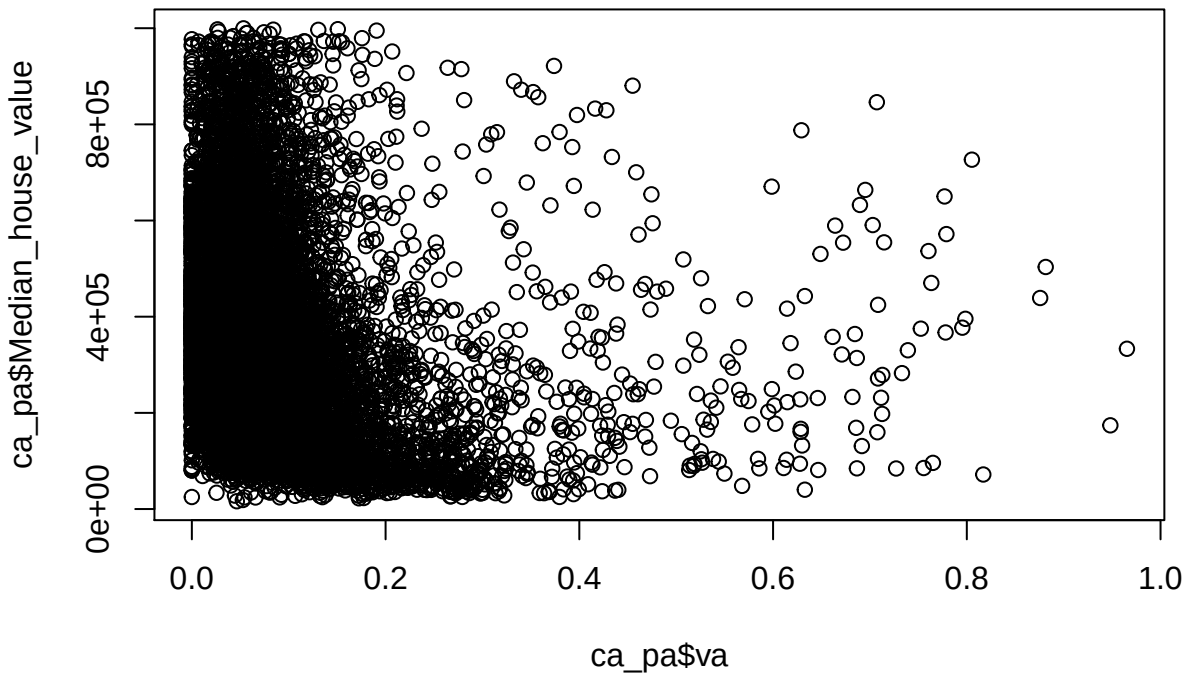
The vacancy rate is the fraction of housing units which are not occupied. The dataframe contains columns giving the total number of housing units for each Census tract, and the number of vacant housing units.

a. Add a new column to the dataframe which contains the vacancy rate. What are the minimum, maximum, mean, and median vacancy rates?

```
ca_pa = ca_pa %>%
  mutate("vacancy rate" = Vacant_units/Total_units)
```

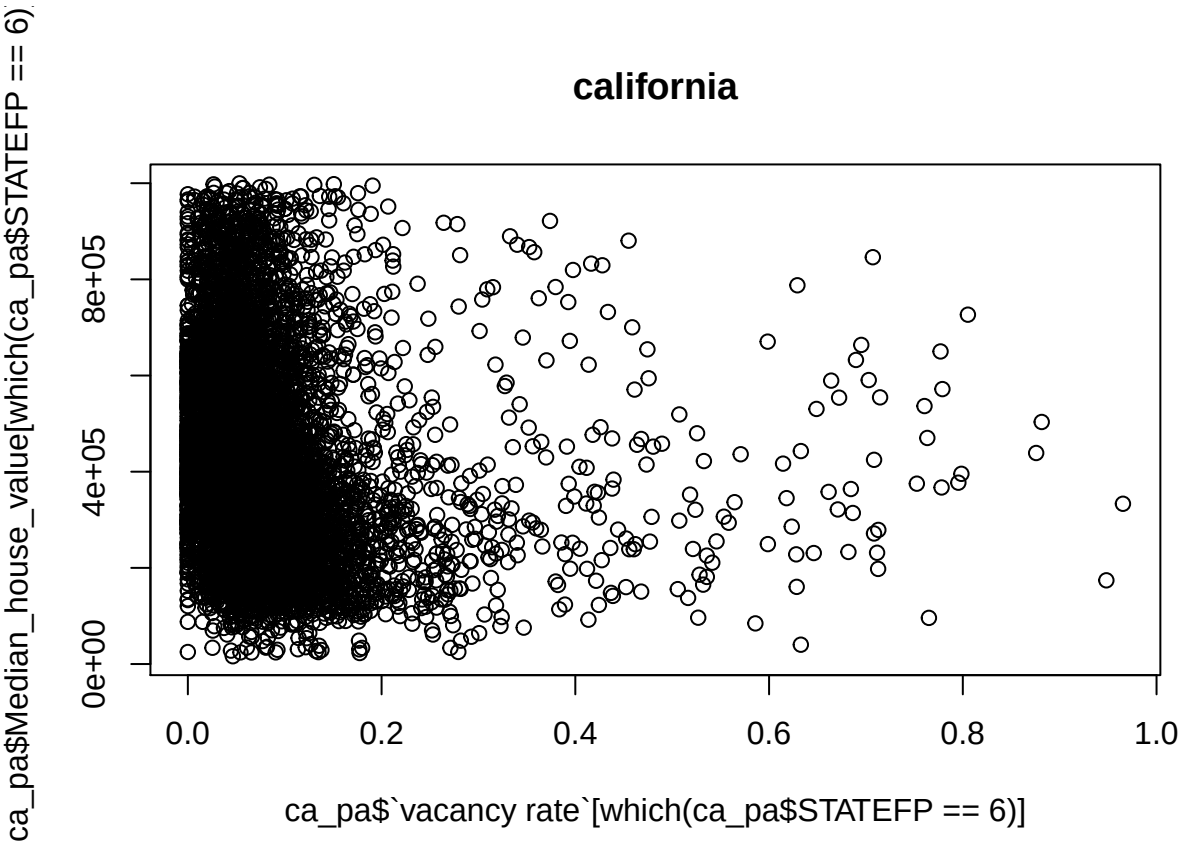
b. Plot the vacancy rate against median house value.

```
plot(x=ca_pa$va,y=ca_pa$Median_house_value)
```

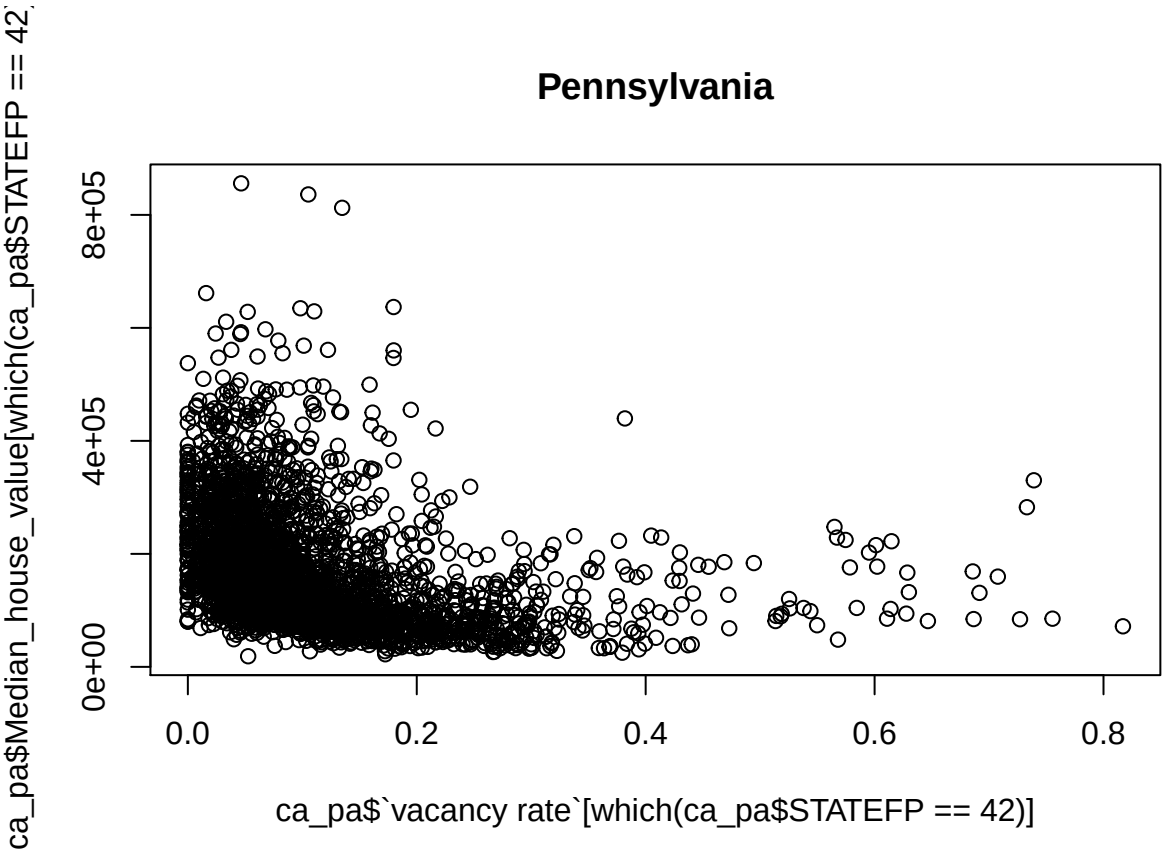


c. Plot vacancy rate against median house value separately for California and for Pennsylvania. Is there a difference?

```
plot(x=ca_pa$`vacancy rate`[which(ca_pa$STATEFP==6)],
     y=ca_pa$Median_house_value[which(ca_pa$STATEFP==6)])
title("california")
```




```
plot(x=ca_pa$`vacancy rate`[which(ca_pa$STATEFP==42)],
     y=ca_pa$Median_house_value[which(ca_pa$STATEFP==42)])
title("Pennsylvania")
```



The houses in Pennsylvania have relatively low price, and there is a more extinct trend that places with high vacancy rate tend to have lower house price.

4. The column `COUNTYFP` contains a numerical code for counties within each state. We are interested in Alameda County (county 1 in California), Santa Clara (county 85 in California), and Allegheny County (county 3 in Pennsylvania).

a. Explain what the block of code at the end of this question is supposed to accomplish, and how it does it. # the code get the median value of median house value in Alameda county. it first get all the observations of Alameda county into vector acca, then find the median value of the median house value in acca.

b. Give a single line of R which gives the same final answer as the block of code. Note: there are at least two ways to do this; you just have to find one.

```
median(ca_pa$Median_house_value[which(ca_pa$STATEFP==6&ca_pa$COUNTYFP==1)])
```

c. For Alameda, Santa Clara and Allegheny Counties, what were the average percentages of housing built since 2005?

```
mean(ca_pa$Built_2005_or_later[which((ca_pa$STATEFP==6&ca_pa$COUNTYFP==1)|
                                     (ca_pa$STATEFP==6&ca_pa$COUNTYFP==85)|
                                     (ca_pa$STATEFP==42&ca_pa$COUNTYFP==3))])
```

d. The `cor` function calculates the correlation coefficient between two variables. What is the correlation between median house value and the percent of housing built since 2005 in (i) the whole data, (ii) all of California, (iii) all of Pennsylvania, (iv) Alameda County, (v) Santa Clara County and (vi) Allegheny County?

```
#(i)
cor(x=ca_pa$Built_2005_or_later,y=ca_pa$Median_house_value)
#(ii)
cor(x=ca_pa$Built_2005_or_later[which(ca_pa$STATEFP==6)],
    y=ca_pa$Median_house_value[which(ca_pa$STATEFP==6)])
#(iii)
cor(x=ca_pa$Built_2005_or_later[which(ca_pa$STATEFP==42)],
    y=ca_pa$Median_house_value[which(ca_pa$STATEFP==42)])
#(iv)
cor(x=ca_pa$Built_2005_or_later[which(ca_pa$STATEFP==6&ca_pa$COUNTYFP==1)],
    y=ca_pa$Median_house_value[which(ca_pa$STATEFP==6&ca_pa$COUNTYFP==1)])
#(v)
cor(x=ca_pa$Built_2005_or_later[which(ca_pa$STATEFP==6&ca_pa$COUNTYFP==85)],
    y=ca_pa$Median_house_value[which(ca_pa$STATEFP==6&ca_pa$COUNTYFP==85)])
#(vi)
cor(x=ca_pa$Built_2005_or_later[which(ca_pa$STATEFP==42&ca_pa$COUNTYFP==3)],
    y=ca_pa$Median_house_value[which(ca_pa$STATEFP==42&ca_pa$COUNTYFP==3)])
```

e. Make three plots, showing median house values against median income, for Alameda, Santa Clara, and Allegheny Counties. (If you can fit the information into one plot, clearly distinguishing the three counties, that's OK too.)

```
ggplot(data = ca_pa[which(ca_pa$STATEFP==6&ca_pa$COUNTYFP==1),]) +
  geom_point(aes(x = Median_household_income[which(ca_pa$STATEFP==6&ca_pa$COUNTYFP==1)], y = Median_hous
  labs(title = "Alameda",       y = "median house value",       x = "median income") +
  theme_bw()
###################################
ggplot(data = ca_pa[which(ca_pa$STATEFP==6&ca_pa$COUNTYFP==85),]) +
  geom_point(aes(x = ca_pa$Median_household_income[which(ca_pa$STATEFP==6&ca_pa$COUNTYFP==85)], y = ca_
  labs(title = "Santa Clara",       y = "median house value",       x = "median income") +
  theme_bw()
###################################
ggplot(data = ca_pa[which(ca_pa$STATEFP==42&ca_pa$COUNTYFP==3),]) +
  geom_point(aes(x = ca_pa$Median_household_income[which(ca_pa$STATEFP==42&ca_pa$COUNTYFP==3)], y = ca_
  labs(title = "and Allegheny",       y = "median house value",       x = "median income") +
  theme_bw()
```

```
acca <- c()
for (tract in 1:nrow(ca_pa)) {
  if (ca_pa$STATEFP[tract] == 6) {
    if (ca_pa$COUNTYFP[tract] == 1) {
      acca <- c(acca, tract)
    }
  }
}
accamhv <- c()
for (tract in acca) {
  accamhv <- c(accamhv, ca_pa[tract,10])
}
median(accamhv)
```

MB.Ch1.11. Run the following code:

```
gender <- factor(c(rep("female", 91), rep("male", 92)))
table(gender)
```

```
## gender
## female   male
##     91     92
```

```
gender <- factor(gender, levels=c("male", "female"))
table(gender)
```

```
## gender
##   male female
##     92     91
```

```
gender <- factor(gender, levels=c("Male", "female"))
# Note the mistake: "Male" should be "male"
table(gender)
```

```
## gender
##   Male female
##      0     91
```

```
table(gender, exclude=NULL)
```

```
## gender
##   Male female   <NA>
##      0     91     92
```

```
rm(gender)  # Remove gender
```

Explain the output from the successive uses of table().

1 factor()is used to encode a vector as a factor, table() list the values and their numbers in the factor 2 the values in the factor is sorted into increasing order of xtable() will list the value according to their order of appearence. 3 “Male” is not in the vector, thus it’s number is 0 4 when forming the factor using “factor(gender, levels=c(”Male“,”female“))”, “male” is excluded automatically, now set excluded=NULL, table() will show a more factor whose name is and number is the number of values excluded before.

MB.Ch1.12. Write a function that calculates the proportion of values in a vector x that exceed some value cutoff.

(a) Use the sequence of numbers 1, 2, . . . , 100 to check that this function gives the result that is expected.

```
cal<-function(x,k){
  if(typeof(x)=="list") total = lengths(x)
  else  total = length(x)

  y = length(which(x>k))
  return(y/total)
}
x = seq(from=1,to=100,by=1)
cal(x,60)
```

```
## [1] 0.4
```

(b) Obtain the vector ex01.36 from the Devore6 (or Devore7) package. These data give the times required for individuals to escape from an oil platform during a drill. Use dotplot() to show the distribution of times. Calculate the proportion of escape times that exceed 7 minutes.

```
library(Devore7)
```

```
## Loading required package: MASS
```

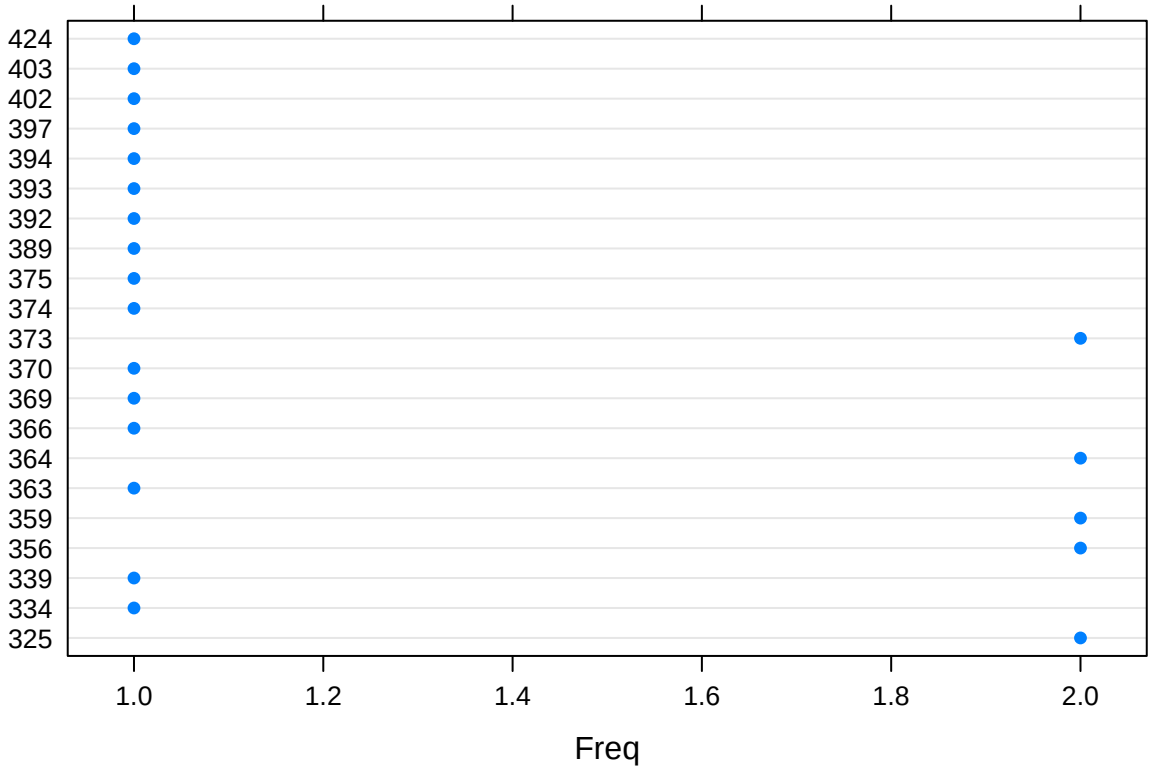
```
##
```

```
## Attaching package: 'MASS'

## The following object is masked from 'package:DAAG':
##
##     hills

## The following object is masked from 'package:dplyr':
##
##     select
```

```
dotplot(ex01.36)
```




```
cal(ex01.36,420)
```

```
##         C1
## 0.03846154
```

MB.Ch1.18. The Rabbit data frame in the MASS library contains blood pressure change measurements on five rabbits (labeled as R1, R2, . . . ,R5) under various control and treatment conditions. Read the help file for more information. Use the unstack() function (three times) to convert Rabbit to the following form:

Treatment Dose R1 R2 R3 R4 R5

1 Control 6.25 0.50 1.00 0.75 1.25 1.5

2 Control 12.50 4.50 1.25 3.00 1.50 1.5

. . . .

```
library(MASS)
Rabbit = Rabbit
```

```
a1 = unstack(Rabbit,BPchange~Animal)
a2 = unstack(Rabbit,Treatment~Animal)
a3 = unstack(Rabbit,Dose~Animal)
data = cbind(a2$R1,a3$R1,a1)
names(data)<-c("Treatment", "Dose", "R1" ,   "R2"   , "R3"  ,  "R4" ,   "R5" )
data
```

```
##    Treatment   Dose    R1    R2    R3    R4   R5
## 1    Control   6.25  0.50  1.00  0.75  1.25  1.5
## 2    Control  12.50  4.50  1.25  3.00  1.50  1.5
## 3    Control  25.00 10.00  4.00  3.00  6.00  5.0
## 4    Control  50.00 26.00 12.00 14.00 19.00 16.0
## 5    Control 100.00 37.00 27.00 22.00 33.00 20.0
## 6    Control 200.00 32.00 29.00 24.00 33.00 18.0
## 7        MDL   6.25  1.25  1.40  0.75  2.60  2.4
## 8        MDL  12.50  0.75  1.70  2.30  1.20  2.5
## 9        MDL  25.00  4.00  1.00  3.00  2.00  1.5
## 10       MDL  50.00  9.00  2.00  5.00  3.00  2.0
## 11       MDL 100.00 25.00 15.00 26.00 11.00  9.0
## 12       MDL 200.00 37.00 28.00 25.00 22.00 19.0
```