

Homework 2 - Report

3170105743 李政达

2020/7/8

The data set `calif_penn_2011.csv` contains information about the housing stock of California and Pennsylvania, as of 2011. Information is 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`.

```
ca_pa <- read.csv("../data/calif_penn_2011.csv", header = TRUE, sep = ",")
```

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

```
rows <- dim(ca_pa)[1]
columns <- dim(ca_pa)[2]
```

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

Ans: this command can figure out the number of missing values in every column.

```
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
```

- 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?

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

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

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

Ans: They are compatible. The command in (c) check the number of missing values in every column, and the command in (e) check the number of rows with incomplete data. We can infer that after purging, the number of missing values in every column will be zero. And we can use the command below to check out the truth.

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

```
## [1] 0
```

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(ca_pa$Median_house_value ~ ca_pa$Built_2005_or_later,
     xlab = "percentage of houses built since 2005",
     ylab = "median house prices")
```



- 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.

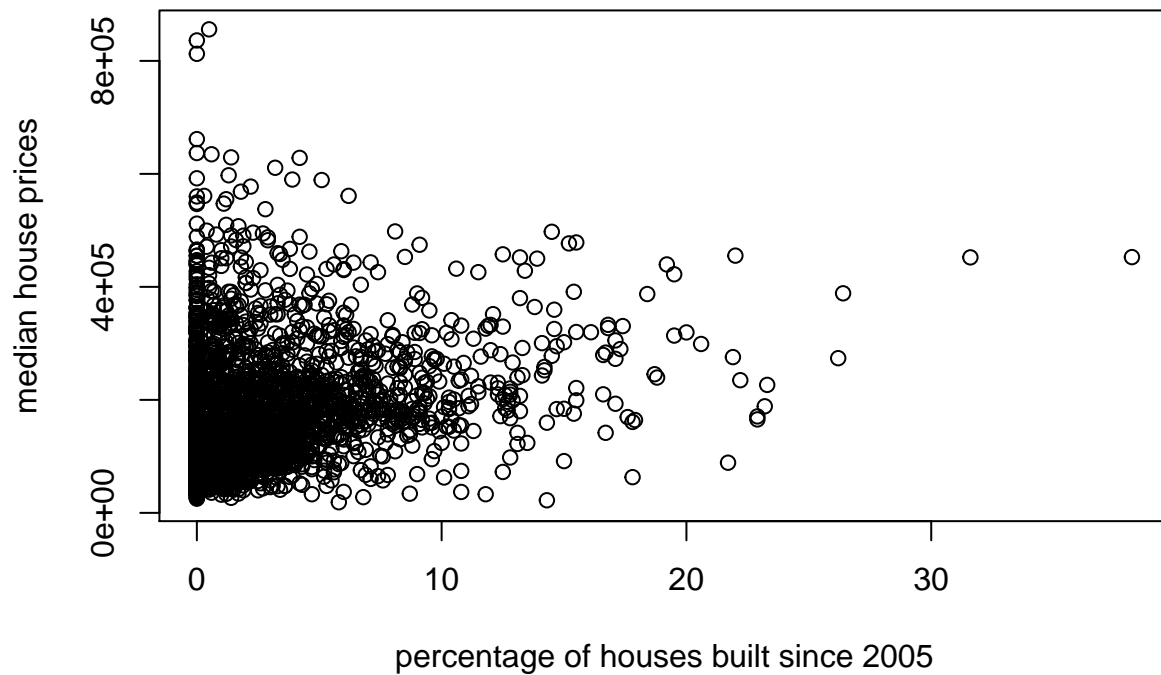
```
plot(ca_pa$Built_2005_or_later[ca_pa$STATEFP == 6],
     ca_pa$Median_house_value[ca_pa$STATEFP == 6],
     xlab = "percentage of houses built since 2005",
     ylab = "median house prices",
     main = "Houses in California")
```

Houses in California



```
plot(ca_pa$Built_2005_or_later[ca_pa$STATEFP == 42],  
     ca_pa$Median_house_value[ca_pa$STATEFP == 42],  
     xlab = "percentage of houses built since 2005",  
     ylab = "median house prices",  
     main = "Houses in Pennsylvania")
```

Houses in 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.

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

```
Vacancy_rate <- ca_pa$Vacant_units / ca_pa$Total_units  
ca_pa <- data.frame(ca_pa, Vacancy_rate)  
max(Vacancy_rate)
```

```
## [1] 0.965311
```

```
min(Vacancy_rate)
```

```
## [1] 0
```

```
mean(Vacancy_rate)
```

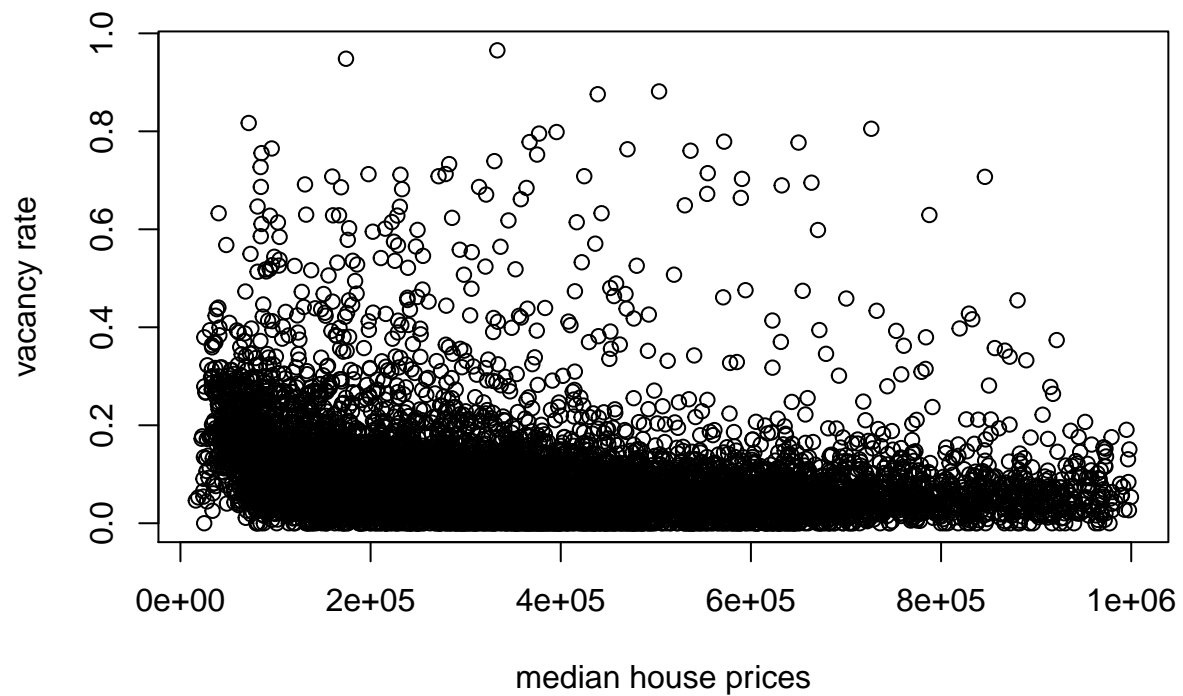
```
## [1] 0.08888789
```

```
median(Vacancy_rate)
```

```
## [1] 0.06767283
```

- Plot the vacancy rate against median house value.

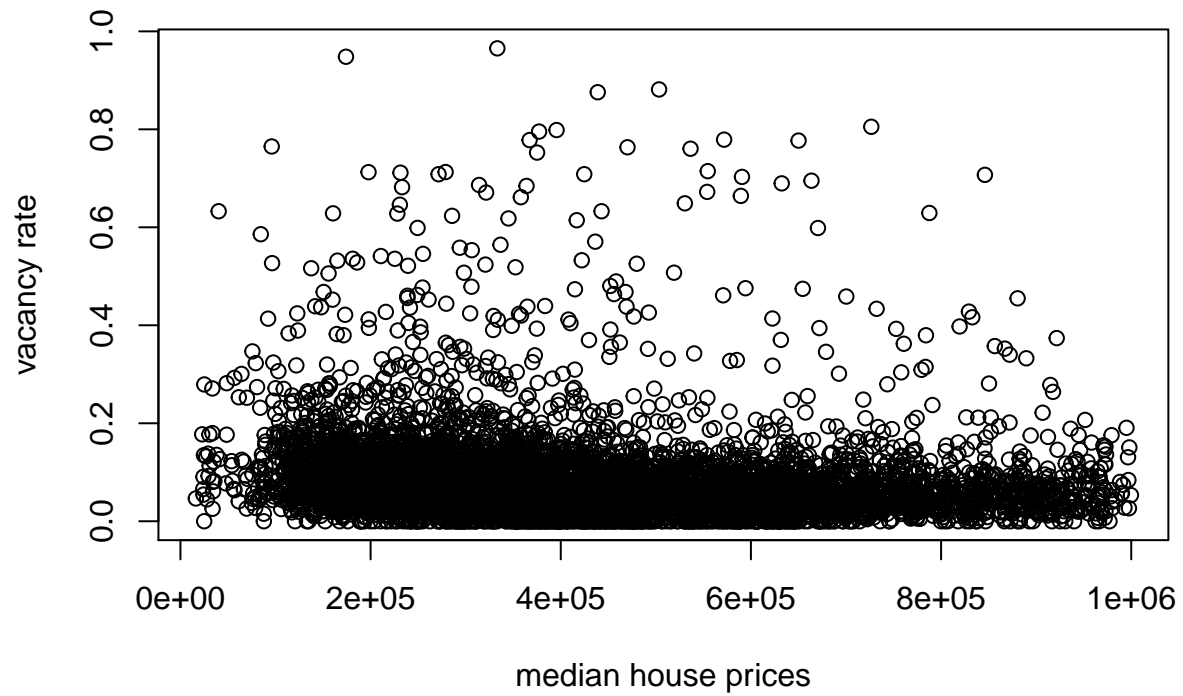
```
plot(ca_pa$Median_house_value, ca_pa$Vacancy_rate,  
     xlab = "median house prices", ylab = "vacancy rate")
```



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

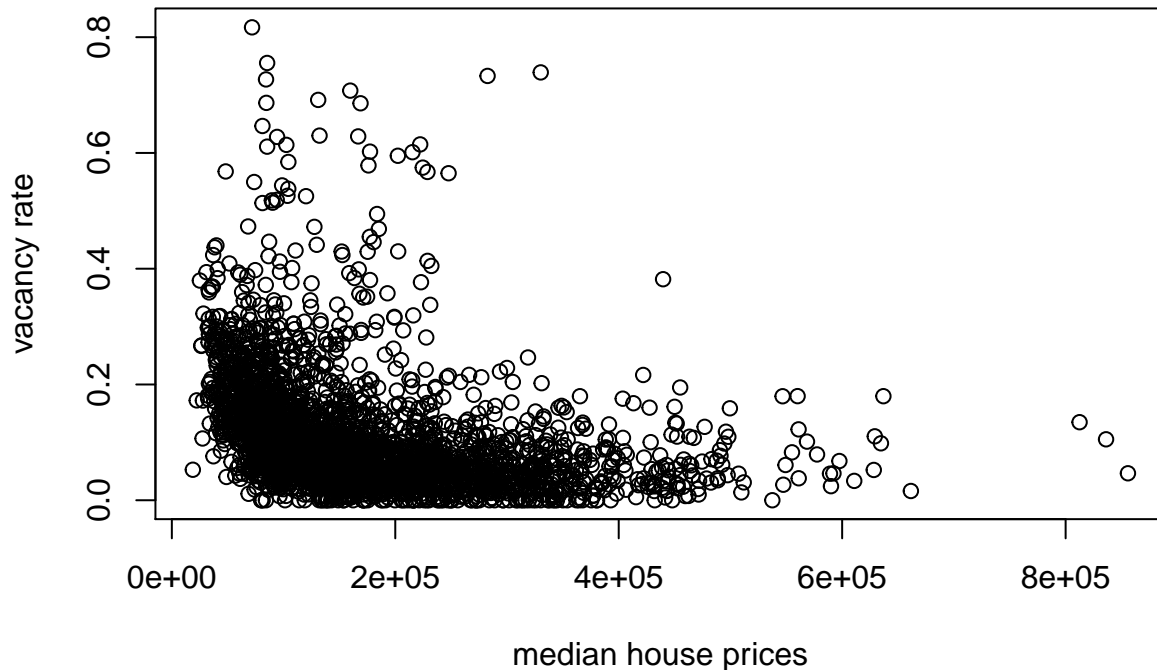
```
plot(ca_pa$Median_house_value[ca_pa$STATEFP == 6],  
     ca_pa$Vacancy_rate[ca_pa$STATEFP == 6],  
     xlab = "median house prices", ylab = "vacancy rate",  
     main = "Houses in California")
```

Houses in California



```
plot(ca_pa$Median_house_value[ca_pa$STATEFP == 42],  
     ca_pa$Vacancy_rate[ca_pa$STATEFP == 42],  
     xlab = "median house prices", ylab = "vacancy rate",  
     main = "Houses in Pennsylvania")
```

Houses in Pennsylvania



The houses in California have higher median house value, and houses with different median house value all have some samples whose vacancy rate is high. The houses in Pennsylvania have lower median house value, and only houses with low median house value have samples whose vacancy rate is high.

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.

Ans: This block of code is supposed to pick up the tracts in Alameda County and compute the median of the median house values of those tracts. The code firstly traverse all the tracts in `ca_pa`, and if a tract matches condition, it will be stored in a new vector `acca`. Then the code traverse again to store the median house values of the tracts in `acca` into `accamhv`. Finally the code call `median` function to compute the median of `accamhv`.

- 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$Median_house_value[ca_pa$STATEFP == 6 & ca_pa$COUNTYFP == 1])
```

```
## [1] 474050
```

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

```
# Alameda
mean(ca_pa$Built_2005_or_later[ca_pa$STATEFP == 6 & ca_pa$COUNTYFP == 1])
```



```
## [1] 2.820468
```

```
# Santa Clara
```

```
mean(ca_pa$Built_2005_or_later[ca_pa$STATEFP == 6 & ca_pa$COUNTYFP == 85])
```

```
## [1] 3.200319
```

```
# Allegheny
```

```
mean(ca_pa$Built_2005_or_later[ca_pa$STATEFP == 42 & ca_pa$COUNTYFP == 3])
```

```
## [1] 1.474219
```

- 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) the whole data
```

```
cor(ca_pa$Median_house_value, ca_pa$Built_2005_or_later)
```

```
## [1] -0.01893186
```

```
# (ii) all of California
```

```
cor(ca_pa$Median_house_value[ca_pa$STATEFP == 6],  
    ca_pa$Built_2005_or_later[ca_pa$STATEFP == 6])
```

```
## [1] -0.1153604
```

```
# (iii) all of Pennsylvania
```

```
cor(ca_pa$Median_house_value[ca_pa$STATEFP == 42],  
    ca_pa$Built_2005_or_later[ca_pa$STATEFP == 42])
```

```
## [1] 0.2681654
```

```
# (iv) Alameda
```

```
cor(ca_pa$Median_house_value[ca_pa$STATEFP == 6 & ca_pa$COUNTYFP == 1],  
    ca_pa$Built_2005_or_later[ca_pa$STATEFP == 6 & ca_pa$COUNTYFP == 1])
```

```
## [1] 0.01303543
```

```
# (v) Santa Clara
```

```
cor(ca_pa$Median_house_value[ca_pa$STATEFP == 6 & ca_pa$COUNTYFP == 85],  
    ca_pa$Built_2005_or_later[ca_pa$STATEFP == 6 & ca_pa$COUNTYFP == 85])
```

```
## [1] -0.1726203
```

```
# (vi) Allegheny
```

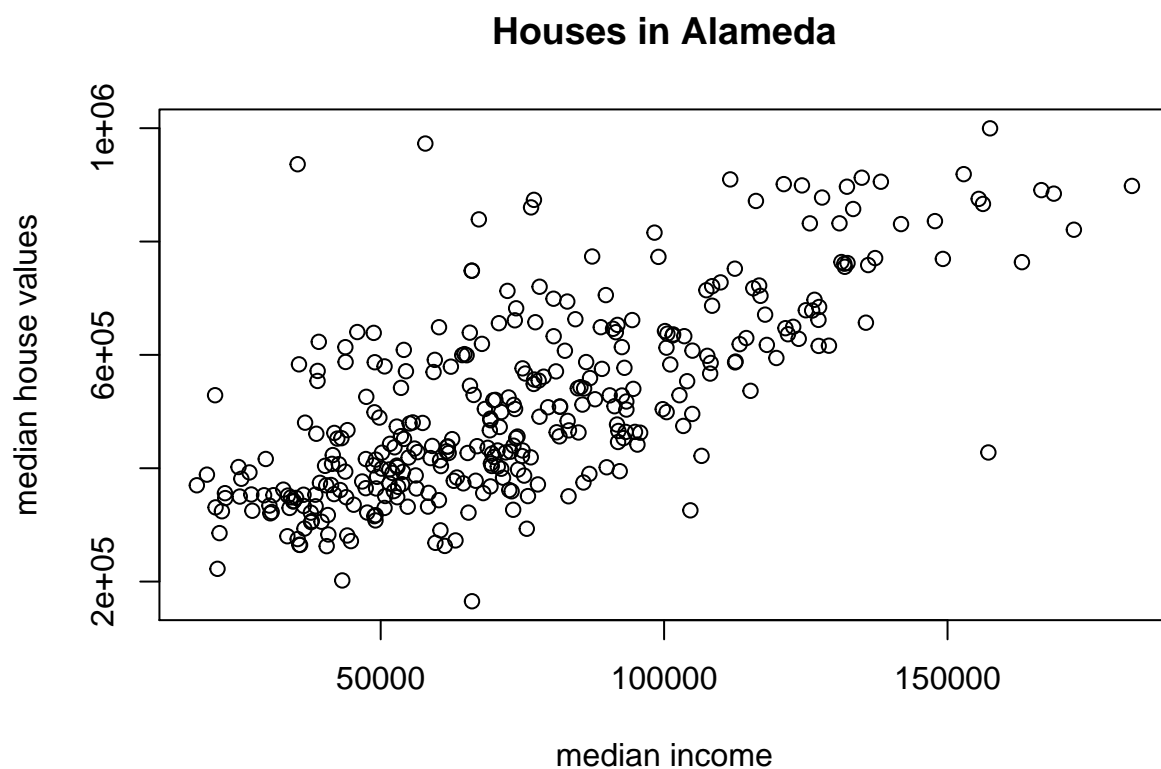
```
cor(ca_pa$Median_house_value[ca_pa$STATEFP == 42 & ca_pa$COUNTYFP == 3],  
    ca_pa$Built_2005_or_later[ca_pa$STATEFP == 42 & ca_pa$COUNTYFP == 3])
```

```
## [1] 0.1939652
```

- 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.)

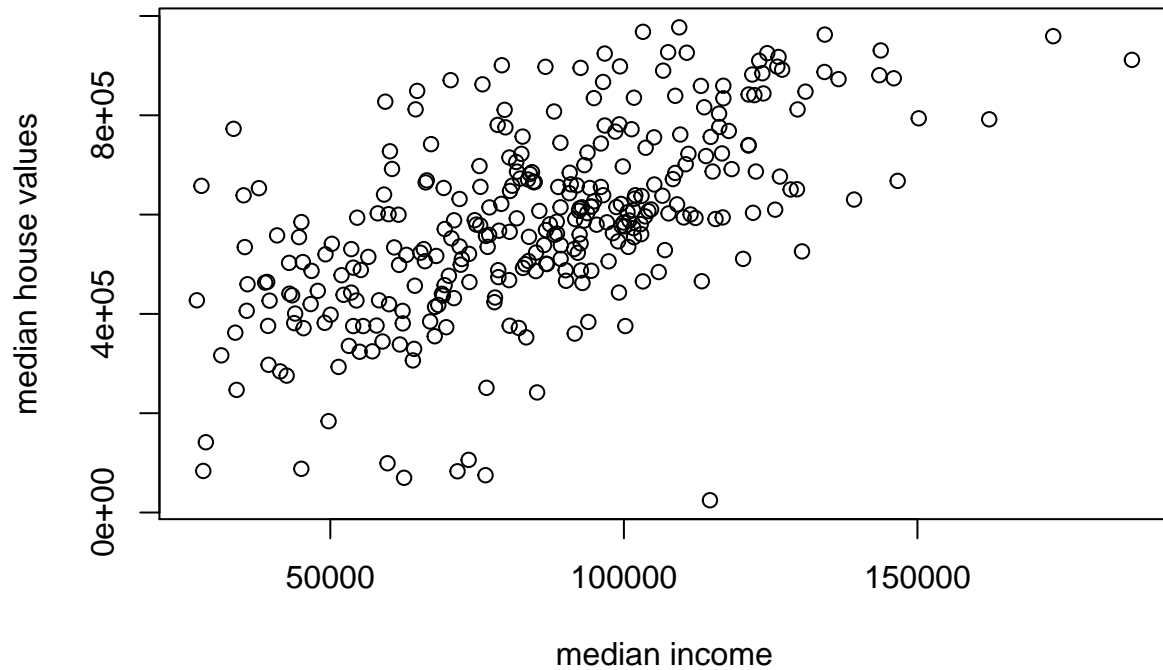
```
#Alameda
```

```
plot(ca_pa$Median_household_income[ca_pa$STATEFP == 6 & ca_pa$COUNTYFP == 1],  
     ca_pa$Median_house_value[ca_pa$STATEFP == 6 & ca_pa$COUNTYFP == 1],  
     xlab = "median income", ylab = "median house values",  
     main = "Houses in Alameda")
```



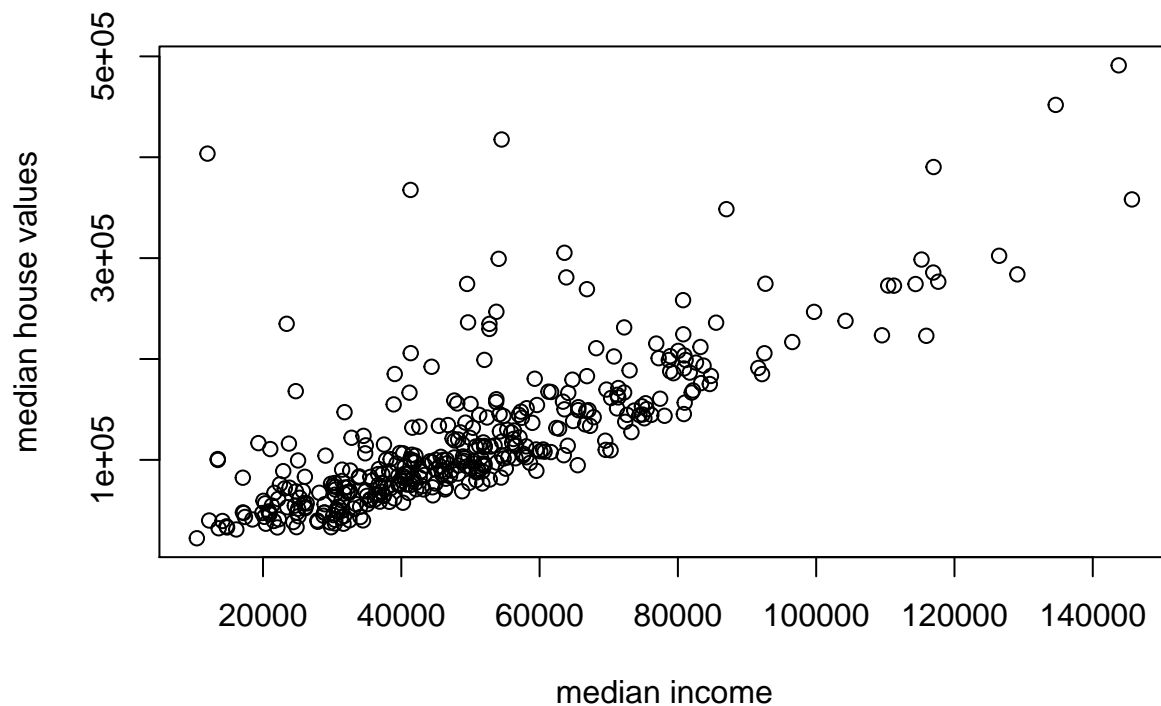
```
# Santa Clara
plot(ca_pa$Median_household_income[ca_pa$STATEFP == 6 & ca_pa$COUNTYFP == 85],
     ca_pa$Median_house_value[ca_pa$STATEFP == 6 & ca_pa$COUNTYFP == 85],
     xlab = "median income", ylab = "median house values",
     main = "Houses in Santa Clara")
```

Houses in Santa Clara



```
# Allegheny
plot(ca_pa$Median_household_income[ca_pa$STATEFP == 42 & ca_pa$COUNTYFP == 3],
     ca_pa$Median_house_value[ca_pa$STATEFP == 42 & ca_pa$COUNTYFP == 3],
     xlab = "median income", ylab = "median house values",
     main = "Houses in Allegheny")
```

Houses in Allegheny



```

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().

Ans:

- We can use `table()` function to display the number of labels in a factor.
- Firstly, `gender` is initialized to be a factor with 91 females and 92 males, so `table(gender)` will display the number of females and the number of males.
- Secondly the code changes the levels of `gender`, exchanges the order of the two labels. So `table(gender)` will display the number of males and the number of females.
- Thirdly, the code sets the levels of `gender` with a wrong case. Because the number of Males is zero, `table(gender)` will display the number of Males, which is zero, and the number of females.
- Finally, the code uses `table(gender, exclude=NULL)` to display all the data, and the data without levels will be showed, too.

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

```
cutoff_proportion <- function(x, cutoff) {
  return(sum(x > cutoff) / length(x))
}
```

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

Ans: We can use the code below to check the correctness. The code will return 0.5 if the code is correct.

```
cutoff_proportion(seq(1, 100), 50)
```

```
## [1] 0.5
```

- (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.

```
# This problem is deleted
```

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

....

```
rabbit <- MASS::Rabbit
treatment <- unstack(rabbit, Treatment ~ Animal)
dose <- unstack(rabbit, Dose ~ Animal)
bpc <- unstack(rabbit, BPchange ~ Animal)
rabbit <- data.frame(treatment[, 1], dose[, 1], bpc)
name <- c("Treatment", "Dose", "R1", "R2", "R3", "R4", "R5")
names(rabbit) <- name
rabbit
```

##	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