# Homework 2: Housing Price

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1. Loading and cleaning

```
a.

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

b. The dataframe has 11275 rows and 34 columns.

```
dim(ca_pa)
```

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

c. The results below are hidden given that it is too long to show them all. apply(ca\_pa,c(1,2),is.na) returns a matrix having the same dimension as ca\_pa, whose elements are Boolean numbers indicating whether the data in ca\_pa is NA. colSums() sums the columns and returns a named vector indicating how many NA elements are there in each column of ca\_pa.

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

d.

```
ca_pa <- na.omit(ca_pa)</pre>
```

e. There are 670 rows containing NA elements, which is now removed.

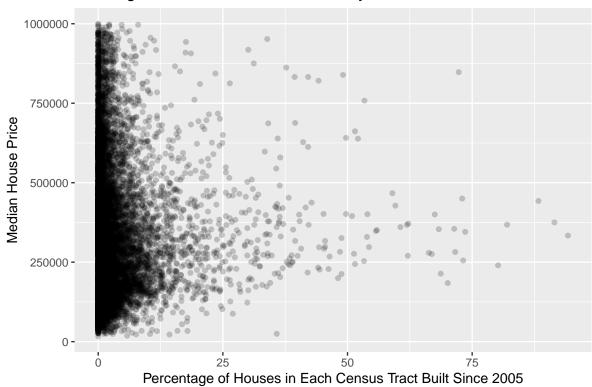
```
11275 - nrow(ca_pa)
```

## [1] 670

- f. The answers in (c) and (e) are compatible. Although we know the number of NA elements in each column, we still have no idea whether they are in the same row or not.
- 2. This Very New House

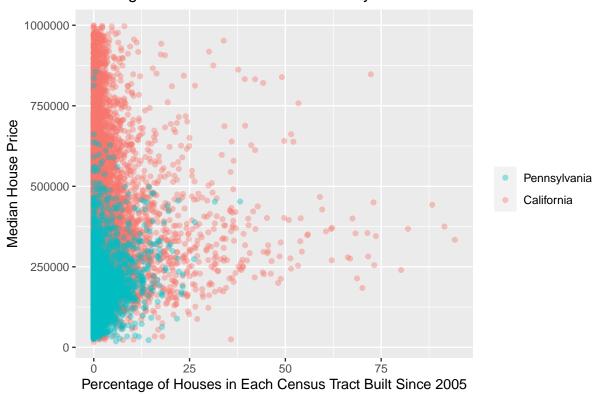
a.

```
library(tidyverse)
ca_pa %>% ggplot(aes(x = Built_2005_or_later, y = Median_house_value)) +
  geom_point(alpha = 0.2) +
labs(x = "Percentage of Houses in Each Census Tract Built Since 2005",
        y = "Median House Price",
        title = "Housing Stock of California and Pennsylvania")
```



reiterlage of Houses in Latif Cerisus Tract Built Since 2003

b.



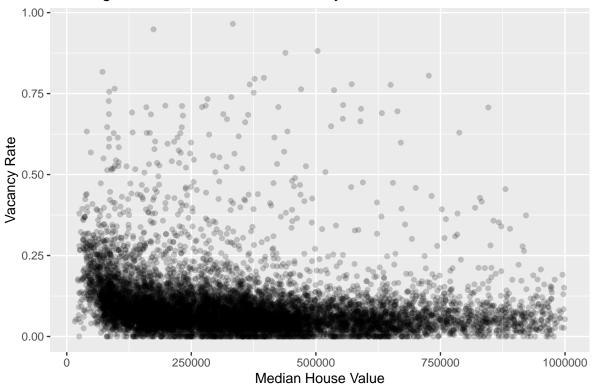
#### 3. Nobody Home

a. The minimum the minimum, maximum, mean, and median vacancy rates are 0.00000, 0.96531, 0.08889 and 0.06767 respectively.

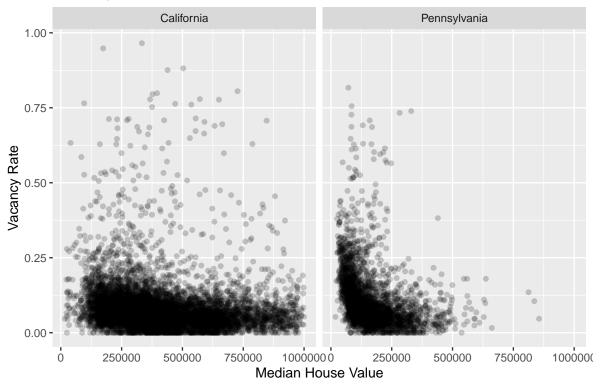
```
ca_pa <- ca_pa %>% mutate(Vacancy_rate = Vacant_units / Total_units)
summary(ca_pa$Vacancy_rate)

## Min. 1st Qu. Median Mean 3rd Qu. Max.
## 0.00000 0.03846 0.06767 0.08889 0.10921 0.96531
b.

ca_pa %>% ggplot(aes(x = Median_house_value, y = Vacancy_rate)) +
    geom_point(alpha = 0.2) +
    labs(x = "Median House Value",
        y = "Vacancy Rate",
        title = "Housing Stock of California and Pennsylvania")
```



c. It is clear that there are fewer Census tracts with high median house value in Pennsylvania, among which there are more Census tracts have higher vacancy rate. Although the number of Census tracts with high median house value in Pennsylvania is not high, their vacancy rate is much lower than those in California. The distribution of vacancy rate among Census tracts in California seems not to change with median house value.

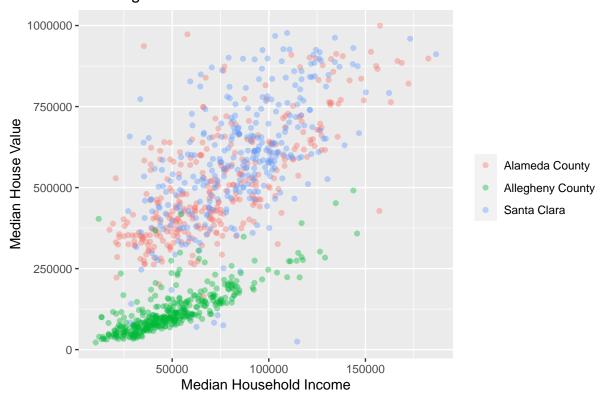


4. a. The first iteration records the row numbers of the county marked as 1 in California to the variable acca. The second iteration records the median house value of the Census tracts in acca to the variable accambv, and finally calculate thier median value (the median value of the median values of the Census tracts recorded in accambv).

```
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)</pre>
accamhv <- c()
for (tract in acca) {
 accamhv <- c(accamhv, ca_pa[tract,10])</pre>
}
median(accamhv)
## [1] 474050
 b.
median(ca_pa[ca_pa$STATEFP == 6 & ca_pa$COUNTYFP == 1, "Median_house_value"])
## [1] 474050
 c.
ca_pa_long <- ca_pa %>%
 gather(key = built_year,
```

```
value = built_numbers,
         dplyr::starts_with('Built'))
ca pa long %>%
 filter((STATEFP == 6 & COUNTYFP %in% c(1,85)) | (STATEFP == 42 & COUNTYFP == 3)) %>%
 group by(COUNTYFP) %>%
 summarise(APHB_2005 = sum((built_year == "Built_2005_or_later") *
                              built_numbers)/sum(built_numbers)) %>%
 ungroup()
## # A tibble: 3 x 2
   COUNTYFP APHB 2005
##
        <int>
                  <dbl>
## 1
           1
                 0.0282
## 2
                 0.0147
           3
## 3
           85
                 0.0320
 d.
p <- ca_pa_long %>% group_by(X) %>%
  summarise(APHB_2005 = sum((built_year == "Built_2005_or_later") *
                              built_numbers)/sum(built_numbers)) %>%
 ungroup()
ca_pa <- dplyr::left_join(ca_pa, p, by = "X")</pre>
rm(p)
cor(ca pa$Median house value, ca pa$APHB 2005)
## [1] -0.01893763
ca pa %>% mutate(STATENAME = ifelse(STATEFP == 6, "California", "Pennsylvania")) %>%
 group_by(STATENAME) %>%
 summarise(Correlation_coefficient = cor(Median_house_value, APHB_2005))
## # A tibble: 2 x 2
   STATENAME Correlation_coefficient
   <chr>
                                    <dh1>
## 1 California
                                   -0.115
## 2 Pennsylvania
                                    0.268
ca pa %>%
 filter((STATEFP == 6 & COUNTYFP %in% c(1,85)) | (STATEFP == 42 & COUNTYFP == 3)) %>%
 mutate(COUNTYNAME = ifelse(COUNTYFP == 1, "Alameda County",
                             ifelse(COUNTYFP == 85, "Santa Clara",
                                    "Allegheny County"))) %>%
 group_by(COUNTYNAME) %>%
 summarise(Correlation_coefficient = cor(Median_house_value, APHB_2005))
## # A tibble: 3 x 2
##
   COUNTYNAME
                      Correlation_coefficient
##
     <chr>
                                        <dbl>
## 1 Alameda County
                                       0.0130
## 2 Allegheny County
                                       0.194
## 3 Santa Clara
                                      -0.173
 e.
ca_pa %>%
 filter((COUNTYFP %in% c(1,85) &STATEFP == 6) | (COUNTYFP==3&STATEFP == 42)) %>%
```

### Housing Stock of 3 Counties



5. (MB.CH1.11) The first line create an variable gender with 2 levels "female" and "male", whose first 91 elements are "female" and the remaining elements are "male". The second line table(gender) shows the factor levels. The third line exchanges the order of levels. The function searches gender first, finding the same level as elements in levels=c("male", "female"), and then changes the numeric order to the new levels. But when it doesn't find the same level, just as line 5, the function removes the old levels and create an new level named Male and matches nothing, thus the result of table(gender) is 0 for level "Male". When NA is included in table, like what line8 does, we can see an NA level with 92 elements, which are exactly those whose level "male" are removed in line5.

```
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)</pre>
```

```
## gender
      male female
         92
  gender <- factor(gender, levels=c("Male", "female"))</pre>
  # Note the mistake: "Male" should be "male"
  table(gender)
  ## gender
  ## Male female
  table(gender, exclude=NULL)
  ## gender
  ## Male female
                     <NA>
  ##
          0
                91
  rm(gender) # Remove gender
6. (MB.CH1.12)
    a.
  cutoff <- function(x, value){</pre>
  prop = sum(x > value) / length(x)
  return(prop)
  cutoff(1:100, 10)
  ## [1] 0.9
  cutoff(1:100, 35)
  ## [1] 0.65
  cutoff(1:100, 35.5)
  ## [1] 0.65
   b.
  library(Devore7)
  dotplot(ex01.36)
```

```
424
403
402
397
394
393
392
389
375
374
373
370
369
366
364
363
359
356
339
334
325
                       1.2
                                      1.4
                                                      1.6
                                                                                    2.0
        1.0
                                                                     1.8
                                             Freq
```

```
cutoff(ex01.36$C1, 420)
```

## [1] 0.03846154

#### 7. (MB.CH1.18)

```
library(MASS)
data(Rabbit)
Treatment <- unstack(Rabbit, Treatment ~ Animal)
Dose <- unstack(Rabbit, Dose ~ Animal)
BPchange <- unstack(Rabbit, BPchange ~ Animal)
Rabbit <- data.frame(Treatment = Treatment[,1],Dose = Dose[,1])
Rabbit <- cbind(Rabbit, BPchange)
Rabbit</pre>
```

```
##
      Treatment
                               R2
                                     RЗ
                                          R4
                                               R5
                 Dose
                         R1
## 1
                 6.25
                       0.50
                            1.00
                                  0.75
                                        1.25
       Control
                                              1.5
## 2
       Control 12.50 4.50
                            1.25
                                  3.00
                                        1.50
## 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
           MDL
                            1.70
## 8
               12.50
                       0.75
                                  2.30
                                        1.20
## 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
```