### Homework 2

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#### 1. Loading and cleaning

(a)

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

(b)

```
dim(ca_pa)
```

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

So the data frame has 11275 rows and 34 columns.

 $(\mathbf{c})$ 

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

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

```
##
                              98
                                                             98
##
                     Bedrooms 2
                                                    Bedrooms_3
##
                              98
##
                     Bedrooms_4
                                           Bedrooms_5_or_more
##
                              98
##
                          Owners
                                                       Renters
##
                             100
##
       Median_household_income
                                        Mean_household_income
##
                             115
```

This command is used to count the number of NA (not a number) values in each column of the data frame.

- The apply() function loops through all the elements of the matrix ca\_pa and applies the is.na() function which returns TRUE if the element is not a number and FALSE otherwise.
- The resulting matrix of TRUE and FALSE values is then given as input to colSums() function, which counts the number of TRUE values in each column.

(d)

```
new_ca_pa<-na.omit(ca_pa)
```

(e)

```
nrow(ca_pa)-nrow(new_ca_pa)
```

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

The result shows that the na.omit() command eliminate 670 rows.

(f) My answer in (c) and (e) are compatible. We can run the following command to verify this:

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

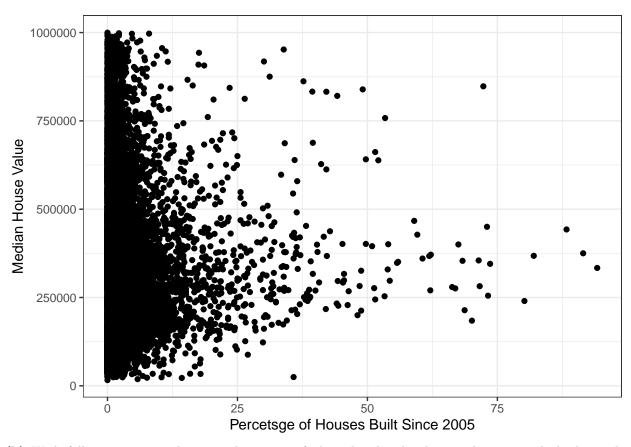
GEO.id2	X	##
0	0	##
COUNTYFP	STATEFP	##
0	0	##
POPULATION	TRACTCE	##
0	0	##
LONGITUDE	LATITUDE	##
0	0	##
Median_house_value	GEO.display.label	##
0	0	##
Vacant_units	Total_units	##
0	0	##
Mean_household_size_owners	Median_rooms	##
0	0	##
Built_2005_or_later	Mean_household_size_renters	##
0	0	##
Built_1990s	Built_2000_to_2004	##
0	0	##

```
Built_1980s
##
                                                     Built_1970s
##
                     Built_1960s
##
                                                     Built_1950s
##
##
                     Built_1940s
                                         Built_1939_or_earlier
##
##
                      Bedrooms 0
                                                      Bedrooms 1
                                0
                                                                0
##
##
                      {\tt Bedrooms\_2}
                                                      Bedrooms_3
##
                                0
                                                                0
                                             Bedrooms_5_or_more
##
                      Bedrooms_4
                                0
##
##
                           Owners
                                                         Renters
##
##
       Median_household_income
                                         {\tt Mean\_household\_income}
##
```

The values of all columns are all 0, which means that we have successfully purged ca\_pa from any row containing NA value.

#### 2. This Very New House

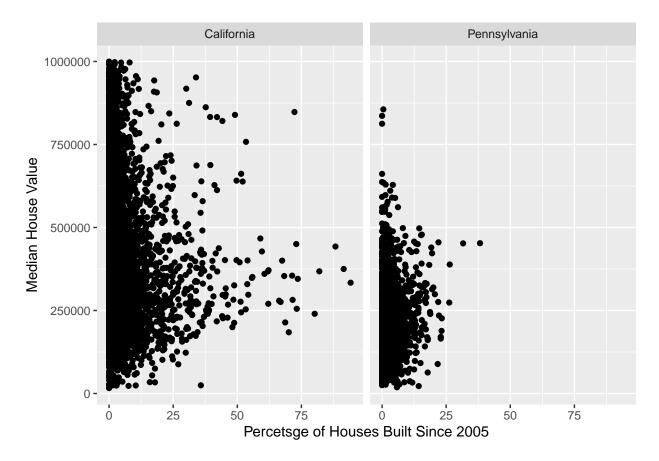
(a)



(b) With following commands, we make a pair of plots that breaks this out by state, which shows the median house prices against Built\_2005\_or\_later in California(state 6) and Pennsylvania(state 42) respectively.

```
d<-data.frame(STATEFP=c(6,42),state=c("California","Pennsylvania"))
ca_pa2<-left_join(new_ca_pa,d,by="STATEFP")</pre>
```

```
ggplot(data = ca_pa2) +
geom_point(aes(x = Built_2005_or_later,
    y = Median_house_value),na.rm = TRUE) +
labs(x = "Percetsge of Houses Built Since 2005",
y = "Median House Value") +facet_wrap(~ state)
```



#### 3. Nobody Home

(a) Add a new column to the dataframe which contains the vacancy rate.

```
ca_pa3 <- ca_pa2 %>%
mutate(Vacant_Rate=Vacant_units/Total_units)
```

Minimum:

```
min(ca_pa3$Vacant_Rate,na.rm = TRUE)
```

**##** [1] 0

Maximum:

```
max(ca_pa3$Vacant_Rate,na.rm = TRUE)
```

## [1] 0.965311

Mean:

```
mean(ca_pa3$Vacant_Rate,na.rm = TRUE)
```

## [1] 0.08888789

Median:

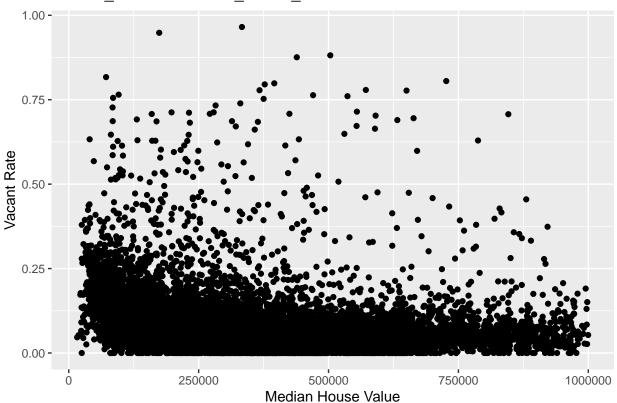
```
median(ca_pa3$Vacant_Rate,na.rm = TRUE)
```

## [1] 0.06767283

(b) Plot the vacancy rate against median house value:

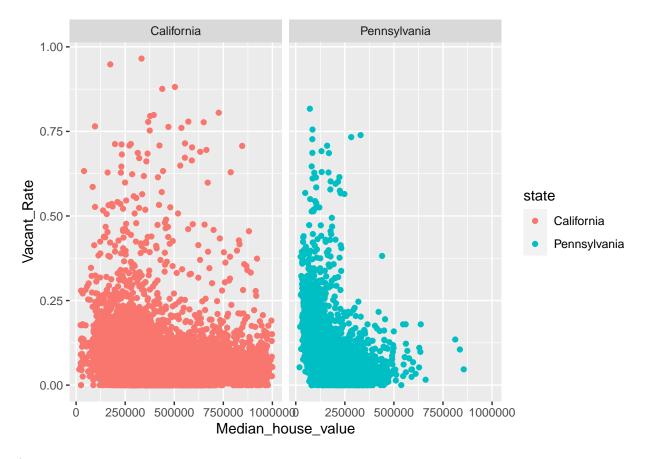
```
ggplot(data = ca_pa3) +
geom_point(aes(x = Median_house_value,
    y = Vacant_Rate),na.rm = TRUE) +
labs(x = "Median House Value",
y = "Vacant Rate",title = "Vacant_Rate vs Median_house_value")
```

### Vacant\_Rate vs Median\_house\_value



(c) Plot vacancy rate against median house value separately for California and for Pennsylvania:

```
ggplot(data = ca_pa3) +
geom_point(aes(x = Median_house_value,
    y = Vacant_Rate,color=state),na.rm = TRUE) +
facet_wrap(~ state)
```



4.

- (a) The block of code is supposed to calculate the median house value in Alameda Country (country 1 in California). It firstly selects the rows of California, whose STATEFP=6, and then selects the rows of Alameda, whose COUNTRYFP=1 among the selected rows. Finally select median value of these rows' Median house value to get the median house value in Alameda.
- (b) We can obtain the same result as the block of code with following command:

```
ca_pa3 %>% filter(STATEFP==6,COUNTYFP==1) %>% {median(.$Median_house_value,na.rm = TRUE)}
```

## [1] 474050

(c) We can obtain the average percentages of housing built since 2005 for Alameda, Santa Clara and Allegheny Counties with following commands: (i) For Alameda:

```
(Alameda_avg <- ca_pa3 %>% filter(STATEFP==6&COUNTYFP==1 )) %>% {mean(.$Built_2005_or_later,na.rm = TRU
```

## [1] 2.820468

(ii) For Santa Clara:

```
(Alameda_avg <- ca_pa3 %>% filter(STATEFP==6&COUNTYFP==85 )) %>% {mean(.$Built_2005_or_later,na.rm = TRI
```

## [1] 3.200319

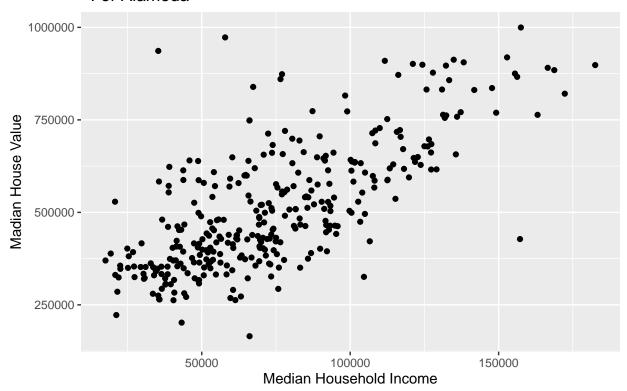
(ii) For Allegheny:

```
(Alameda_avg <- ca_pa3 %>% filter(STATEFP==42&COUNTYFP==3 )) %>% {mean(.$Built_2005_or_later,na.rm = TR
## [1] 1.474219
(d) the correlation between median house value and the percent of housing built since 2005 (i) In the whole
data:
cor(ca_pa3$Median_house_value,ca_pa3$Built_2005_or_later)
## [1] -0.01893186
 (ii) In all of California
ca_pa3 %>% filter(STATEFP==6) %>%
 {cor(.$Median_house_value,.$Built_2005_or_later)}
## [1] -0.1153604
(iii) In all of Pennsylvania
ca_pa3 %>% filter(STATEFP==42) %>%
 {cor(.$Median_house_value,.$Built_2005_or_later)}
## [1] 0.2681654
(iv) In Alameda County
ca_pa3 %>% filter(STATEFP==6,COUNTYFP==1) %>%
  {cor(.$Median_house_value,.$Built_2005_or_later)}
## [1] 0.01303543
 (v) In Santa Clara County
ca pa3 %>% filter(STATEFP==6,COUNTYFP==85) %>%
  {cor(.$Median_house_value,.$Built_2005_or_later)}
## [1] -0.1726203
(vi) In Allegheny Count
ca_pa3 %>% filter(STATEFP==42,COUNTYFP==3) %>%
 {cor(.$Median_house_value,.$Built_2005_or_later)}
## [1] 0.1939652
```

(e) Median house values against median income. (i) For Alameda:

```
ca_pa3 %>% filter(STATEFP==6,COUNTYFP==1) %>%
ggplot() +
geom_point(aes(x = Median_household_income,
    y = Median_house_value),na.rm = TRUE) +
labs(x = "Median Household Income",
y = "Madian House Value",title = "Median House Values vs Median Income
For Alameda")
```

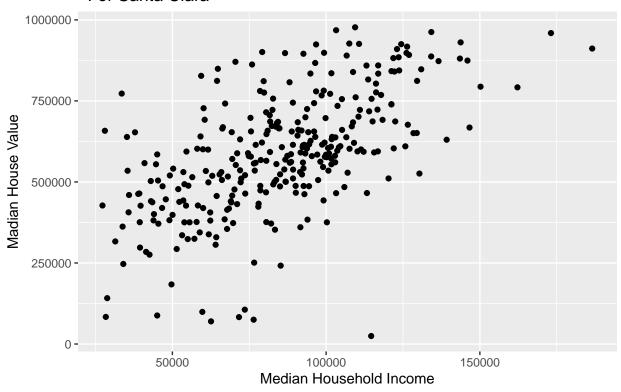
### Median House Values vs Median Income For Alameda



#### (ii) For Santa Clara:

```
ca_pa3 %>% filter(STATEFP==6,COUNTYFP==85) %>%
ggplot() +
geom_point(aes(x = Median_household_income,
    y = Median_house_value),na.rm = TRUE) +
labs(x = "Median Household Income",
y = "Madian House Value",title = "Median House Values vs Median Income
For Santa Clara")
```

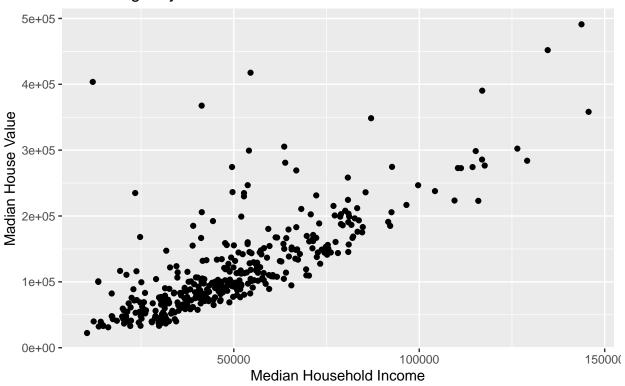
## Median House Values vs Median Income For Santa Clara



#### (iii) For Allegheny:

```
ca_pa3 %>% filter(STATEFP==42,COUNTYFP==3) %>%
ggplot() +
geom_point(aes(x = Median_household_income,
    y = Median_house_value),na.rm = TRUE) +
labs(x = "Median Household Income",
y = "Madian House Value",title = "Median House Values vs Median Income
For Allegheny")
```

# Median House Values vs Median Income For Allegheny



#### MB.Ch1.11. Run the following code:

```
gender <- factor(c(rep("female", 91), rep("male", 92)))</pre>
table(gender)
## gender
             male
## female
##
       91
               92
Explain: table() gives the frequency of each element "female" and "male".
gender <- factor(gender, levels=c("male", "female"))</pre>
table(gender)
## gender
##
     male female
       92
##
```

Explain: the output gives the frequency in levels, that is "male" and "female".

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

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

Male female

91

<NA>

92

Explain: the output gives the frequency of levels, that is "Male" and "female". While there is no "Male" in gender, so the value of Male is 0.

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

Explain: The argument "exclude=NULL" require to report NA's, so the result gives =92 with respond to "male" that is not in the levels.

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

#### MB.Ch1.12.

##

##

```
k<-0
Func <- function(x,cutoff_value){
  for(i in x){
    if (i>cutoff_value){
      k<-k+1
    }
  }
  prop <- k/length(x)
  return(prop)
}</pre>
```

(a) For the sequence of numbers 1, 2, . . . , 100, we set the value cutoff 35,67.3 and 89, then the expected proportion result is 0.65, 0.33 and 0.11 correspondingly. And following codes check this.

```
x<-1:100
vc<-35
Func(x,vc)
```

## [1] 0.65

```
vc<-67.3
Func(x,vc)
```

## [1] 0.33

```
vc<-89
Func(x,vc)
```

## [1] 0.11

MB.Ch1.18. Using following commands, we can convert Rabbit to the required form.

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

```
##
     Treatment
                                              R5
                 Dose
                        R1
                              R2
                                    R3
                                         R4
## 1
                 6.25 0.50 1.00 0.75
       Control
                                       1.25
                                             1.5
       Control 12.50 4.50 1.25 3.00
## 2
                                       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
           \mathtt{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
```