

Mini Project 1

Firstly, we load our data.

```
## set options
options(scipen = 100)
```

```
## load packages
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
```

```
library(psych)
library(pracma)
```

```
##
## Attaching package: 'pracma'
```

```
## The following objects are masked from 'package:psych':
##
##   logit, polar
```

```
library(ggplot2)
```

```
##
## Attaching package: 'ggplot2'
```

```
## The following objects are masked from 'package:psych':
##
##   %+%, alpha
```

```
# load data
dat <- readRDS("house_data.RDS")

# check data
ls(dat)
```

```
##   [1] "AdjSalePrice"  "Bathrooms"    "Bedrooms"     "BldgGrade"
##   [5] "DocumentDate" "ID"            "ImpsVal"       "LandVal"
##   [9] "NbrLivingUnits" "NewConstruction" "PropertyID"    "PropertyType"
##  [13] "SalePrice"     "SqFtFinBasement" "SqFtLot"       "SqFtTotLiving"
##  [17] "TrafficNoise"  "ym"            "YrBuilt"       "YrRenovated"
##  [21] "zhvi_idx"      "zhvi_px"       "ZipCode"
```

We found there are 23 variables in this data set.

And now we are going to do some data tidy, including these steps:

- drop all the non-numeric variables
- drop ID and Party ID
- drop the variable with low variance

```
dat <- dat[,sapply(dat, is.numeric)] # drop non-numeric
dat <- dat %>% # drop ID and party ID
  select(!c(ID, PropertyID))
dat %>% # drop low variance variable
  summarise(across(everything(), ~var(scale(., center = FALSE), na.rm = TRUE)))
```

```
## SalePrice zhvi_px zhvi_idx AdjSalePrice NbrLivingUnits SqFtLot
## 1 0.3177451 0.008517194 0.008517194 0.3173574 0.02399549 0.8591889
## SqFtTotLiving SqFtFinBasement Bathrooms Bedrooms BldgGrade YrBuilt
## 1 0.161738 0.6919151 0.1107266 0.0672611 0.02307366 0.0002364283
## YrRenovated TrafficNoise LandVal ImpsVal ZipCode
## 1 0.9488276 0.8788141 0.408023 0.3624847 0.0000002848018
```

Then we create a correlate coefficient matrix to check the relationship between different variables.

```
lowerCor(dat)
```

```
##          SlPrc zhv_p zhv_d AdjSP NbrLU SqFtL SqFTL SqFFB Bthrm Bdrms
## SalePrice      1.00
## zhvi_px        0.10 1.00
## zhvi_idx        0.10 1.00 1.00
## AdjSalePrice    0.99 -0.04 -0.04 1.00
## NbrLivingUnits  0.02 0.02 0.02 0.02 1.00
## SqFtLot         0.13 -0.01 -0.01 0.14 -0.01 1.00
## SqFtTotLiving   0.69 -0.02 -0.02 0.70 0.06 0.20 1.00
## SqFtFinBasement 0.30 -0.01 -0.01 0.30 0.07 0.04 0.41 1.00
## Bathrooms       0.52 -0.03 -0.03 0.53 0.11 0.11 0.76 0.27 1.00
## Bedrooms        0.31 0.00 0.00 0.31 0.17 0.07 0.60 0.31 0.54 1.00
## BldgGrade        0.67 -0.05 -0.05 0.68 -0.05 0.15 0.77 0.14 0.66 0.37
## YrBuilt          0.08 -0.02 -0.02 0.08 -0.11 0.07 0.31 -0.21 0.47 0.15
## YrRenovated      0.11 0.00 0.00 0.11 0.01 0.01 0.07 0.11 0.07 0.05
## TrafficNoise     -0.01 0.01 0.01 -0.01 0.08 -0.01 -0.06 0.05 -0.06 -0.04
## LandVal          0.80 -0.02 -0.02 0.81 0.03 0.07 0.47 0.28 0.31 0.20
## ImpsVal          0.83 -0.03 -0.03 0.83 0.00 0.11 0.76 0.28 0.63 0.35
## ZipCode          -0.07 0.00 0.00 -0.07 0.06 -0.12 -0.20 0.14 -0.20 -0.15
## BldgG
## SalePrice
## zhvi_px
## zhvi_idx
## AdjSalePrice
## NbrLivingUnits
## SqFtLot
## SqFtTotLiving
## SqFtFinBasement
## Bathrooms
## Bedrooms
## BldgGrade      1.00
## YrBuilt        0.43
## YrRenovated    0.03
## TrafficNoise   -0.06
## LandVal        0.46
## ImpsVal        0.75
## ZipCode        -0.19
## YrBlt YrRnv TrffN LndVl ImpsV ZipCd
## YrBuilt      1.00
## YrRenovated  -0.25 1.00
## TrafficNoise -0.15 0.02 1.00
## LandVal      -0.11 0.13 0.00 1.00
## ImpsVal       0.18 0.12 -0.01 0.57 1.00
## ZipCode      -0.38 0.08 0.10 -0.07 -0.03 1.00
```

Certificate if the variance and corr variance is equal. This step aims to make sure if the data is correct.

```
var(dat, na.rm = TRUE) == cov(dat, use = "complete.obs")
```

```
## SalePrice zhvi_px zhvi_idx AdjSalePrice NbrLivingUnits SqFtLot
## SalePrice TRUE TRUE TRUE TRUE TRUE TRUE
## zhvi_px TRUE TRUE TRUE TRUE TRUE TRUE
## zhvi_idx TRUE TRUE TRUE TRUE TRUE TRUE
## AdjSalePrice TRUE TRUE TRUE TRUE TRUE TRUE
## NbrLivingUnits TRUE TRUE TRUE TRUE TRUE TRUE
## SqFtLot TRUE TRUE TRUE TRUE TRUE TRUE
## SqFtTotLiving TRUE TRUE TRUE TRUE TRUE TRUE
## SqFtFinBasement TRUE TRUE TRUE TRUE TRUE TRUE
## Bathrooms TRUE TRUE TRUE TRUE TRUE TRUE
## Bedrooms TRUE TRUE TRUE TRUE TRUE TRUE
## BldgGrade TRUE TRUE TRUE TRUE TRUE TRUE
## YrBuilt TRUE TRUE TRUE TRUE TRUE TRUE
## YrRenovated TRUE TRUE TRUE TRUE TRUE TRUE
## TrafficNoise TRUE TRUE TRUE TRUE TRUE TRUE
## LandVal TRUE TRUE TRUE TRUE TRUE TRUE
## ImpsVal TRUE TRUE TRUE TRUE TRUE TRUE
## ZipCode TRUE TRUE TRUE TRUE TRUE TRUE
## SqFtTotLiving SqFtFinBasement Bathrooms Bedrooms BldgGrade
## SalePrice TRUE TRUE TRUE TRUE TRUE
## zhvi_px TRUE TRUE TRUE TRUE TRUE
## zhvi_idx TRUE TRUE TRUE TRUE TRUE
## AdjSalePrice TRUE TRUE TRUE TRUE TRUE
## NbrLivingUnits TRUE TRUE TRUE TRUE TRUE
## SqFtLot TRUE TRUE TRUE TRUE TRUE
## SqFtTotLiving TRUE TRUE TRUE TRUE TRUE
## SqFtFinBasement TRUE TRUE TRUE TRUE TRUE
## Bathrooms TRUE TRUE TRUE TRUE TRUE
## Bedrooms TRUE TRUE TRUE TRUE TRUE
## BldgGrade TRUE TRUE TRUE TRUE TRUE
## YrBuilt TRUE TRUE TRUE TRUE TRUE
## YrRenovated TRUE TRUE TRUE TRUE TRUE
## TrafficNoise TRUE TRUE TRUE TRUE TRUE
## LandVal TRUE TRUE TRUE TRUE TRUE
## ImpsVal TRUE TRUE TRUE TRUE TRUE
## ZipCode TRUE TRUE TRUE TRUE TRUE
## YrBuilt YrRenovated TrafficNoise LandVal ImpsVal ZipCode
## SalePrice TRUE TRUE TRUE TRUE TRUE TRUE
## zhvi_px TRUE TRUE TRUE TRUE TRUE TRUE
## zhvi_idx TRUE TRUE TRUE TRUE TRUE TRUE
## AdjSalePrice TRUE TRUE TRUE TRUE TRUE TRUE
## NbrLivingUnits TRUE TRUE TRUE TRUE TRUE TRUE
## SqFtLot TRUE TRUE TRUE TRUE TRUE TRUE
## SqFtTotLiving TRUE TRUE TRUE TRUE TRUE TRUE
## SqFtFinBasement TRUE TRUE TRUE TRUE TRUE TRUE
## Bathrooms TRUE TRUE TRUE TRUE TRUE TRUE
## Bedrooms TRUE TRUE TRUE TRUE TRUE TRUE
## BldgGrade TRUE TRUE TRUE TRUE TRUE TRUE
## YrBuilt TRUE TRUE TRUE TRUE TRUE TRUE
## YrRenovated TRUE TRUE TRUE TRUE TRUE TRUE
## TrafficNoise TRUE TRUE TRUE TRUE TRUE TRUE
## LandVal TRUE TRUE TRUE TRUE TRUE TRUE
## ImpsVal TRUE TRUE TRUE TRUE TRUE TRUE
## ZipCode TRUE TRUE TRUE TRUE TRUE TRUE
```

```
Mcov <- var(dat, na.rm = TRUE)

# if we centre the data, we get the same result
c_dat <- scale(dat, center = TRUE, scale = FALSE)
Mcov_c <- var(c_dat, na.rm = TRUE)

near(Mcov_c, Mcov) # covariance matrix is same
```

##	SalePrice	zhvi_px	zhvi_idx	AdjSalePrice	NbrLivingUnits	SqFtLot
##	SalePrice	FALSE	FALSE	TRUE	FALSE	TRUE
##	zhvi_px	FALSE	FALSE	TRUE	FALSE	TRUE
##	zhvi_idx	TRUE	TRUE	TRUE	TRUE	TRUE
##	AdjSalePrice	FALSE	FALSE	TRUE	FALSE	TRUE
##	NbrLivingUnits	TRUE	TRUE	TRUE	TRUE	TRUE
##	SqFtLot	FALSE	FALSE	TRUE	FALSE	TRUE
##	SqFtTotLiving	TRUE	TRUE	TRUE	FALSE	TRUE
##	SqFtFinBasement	FALSE	TRUE	TRUE	FALSE	TRUE
##	Bathrooms	TRUE	TRUE	TRUE	TRUE	TRUE
##	Bedrooms	TRUE	TRUE	TRUE	TRUE	TRUE
##	BldgGrade	TRUE	TRUE	TRUE	TRUE	TRUE
##	YrBuilt	TRUE	TRUE	TRUE	TRUE	TRUE
##	YrRenovated	TRUE	TRUE	TRUE	TRUE	TRUE
##	TrafficNoise	TRUE	TRUE	TRUE	TRUE	TRUE
##	LandVal	TRUE	FALSE	TRUE	TRUE	FALSE
##	ImpsVal	TRUE	FALSE	TRUE	TRUE	FALSE
##	ZipCode	TRUE	TRUE	TRUE	TRUE	TRUE
##	SqFtTotLiving	SqFtFinBasement	Bathrooms	Bedrooms	BldgGrade	
##	SalePrice	TRUE	FALSE	TRUE	TRUE	TRUE
##	zhvi_px	TRUE	TRUE	TRUE	TRUE	TRUE
##	zhvi_idx	TRUE	TRUE	TRUE	TRUE	TRUE
##	AdjSalePrice	FALSE	FALSE	TRUE	TRUE	TRUE
##	NbrLivingUnits	TRUE	TRUE	TRUE	TRUE	TRUE
##	SqFtLot	TRUE	TRUE	TRUE	TRUE	TRUE
##	SqFtTotLiving	TRUE	TRUE	TRUE	TRUE	TRUE
##	SqFtFinBasement	TRUE	TRUE	TRUE	TRUE	TRUE
##	Bathrooms	TRUE	TRUE	TRUE	TRUE	TRUE
##	Bedrooms	TRUE	TRUE	TRUE	TRUE	TRUE
##	BldgGrade	TRUE	TRUE	TRUE	TRUE	TRUE
##	YrBuilt	TRUE	TRUE	TRUE	TRUE	TRUE
##	YrRenovated	TRUE	TRUE	TRUE	TRUE	TRUE
##	TrafficNoise	TRUE	TRUE	TRUE	TRUE	TRUE
##	LandVal	FALSE	FALSE	TRUE	TRUE	TRUE
##	ImpsVal	FALSE	FALSE	TRUE	TRUE	TRUE
##	ZipCode	TRUE	TRUE	TRUE	TRUE	TRUE
##	YrBuilt	YrRenovated	TrafficNoise	LandVal	ImpsVal	ZipCode
##	SalePrice	TRUE	TRUE	TRUE	TRUE	TRUE
##	zhvi_px	TRUE	TRUE	TRUE	FALSE	FALSE
##	zhvi_idx	TRUE	TRUE	TRUE	TRUE	TRUE
##	AdjSalePrice	TRUE	TRUE	TRUE	TRUE	TRUE
##	NbrLivingUnits	TRUE	TRUE	TRUE	TRUE	TRUE
##	SqFtLot	TRUE	TRUE	TRUE	FALSE	FALSE
##	SqFtTotLiving	TRUE	TRUE	TRUE	FALSE	FALSE
##	SqFtFinBasement	TRUE	TRUE	TRUE	FALSE	FALSE
##	Bathrooms	TRUE	TRUE	TRUE	TRUE	TRUE
##	Bedrooms	TRUE	TRUE	TRUE	TRUE	TRUE
##	BldgGrade	TRUE	TRUE	TRUE	TRUE	TRUE
##	YrBuilt	TRUE	TRUE	TRUE	TRUE	TRUE
##	YrRenovated	TRUE	TRUE	TRUE	FALSE	TRUE
##	TrafficNoise	TRUE	TRUE	TRUE	TRUE	TRUE
##	LandVal	TRUE	FALSE	TRUE	FALSE	FALSE
##	ImpsVal	TRUE	TRUE	TRUE	FALSE	FALSE
##	ZipCode	TRUE	TRUE	TRUE	TRUE	TRUE

Then we perform the eigen-decomposition of the covariance matrix and check the results. This step can divided a matrix into a eigen values and eigen vectors.

```
eigMcov <- eigen(Mcov)
eigMcov
```

```
## eigen() decomposition
## $values
## [1] 326208688074.6403808593750 18235914108.6289024353027
## [3] 7213699330.8674087524414 2907594354.4751238822937
## [5] 819768275.0459656715393 265647813.2387284040451
## [7] 358827.5509719259571 193944.6326406990411
## [9] 139701.7964463515382 2380.4086192770487
## [11] 540.1920212516691 0.5116549668237
## [13] 0.3719691774605 0.2991274858044
## [15] 0.1831829428816 0.0231452300706
## [17] -0.0000002413329
##
## $vectors
## [1,] [2,] [3,] [4,]
## [1,] -0.602048190758632 -0.08632353123778 -0.348452357792817 -0.54053693604472
## [2,] -0.000564338517070 -0.00950159350215 -0.089000806375808 -0.61488790468656
## [3,] -0.000000001296734 -0.00000002183270 -0.000000204505529 -0.00000141288581
## [4,] -0.671371692965639 -0.08286364811084 -0.239609377807863 0.56099741735375
## [5,] -0.000000006144448 -0.00000003859377 -0.000000008546372 -0.00000002763985
## [6,] -0.006723293476451 0.00466043018045 -0.030830105151754 0.02224672094794
## [7,] -0.001135948383779 0.00202124256211 0.000533277664980 0.00007708117987
## [8,] -0.000234642301026 0.00000292508448 0.000491919926047 -0.00007472283429
## [9,] -0.000000729351419 0.00000185276710 0.000000243518647 0.00000013819191
## [10,] -0.000000506493967 0.00000099188393 0.000000216660954 -0.00000014412608
## [11,] -0.000001427115810 0.00000261541785 0.000001065153434 0.00000047833112
## [12,] -0.000004410954214 0.00006509561867 -0.000045394841596 0.00001309590428
## [13,] -0.000088015559774 -0.00001345352307 0.000496762583082 -0.00011173299429
## [14,] 0.000000010813782 -0.00000005661127 0.000000127913354 -0.00000009564167
## [15,] -0.263765414693926 -0.60932551094283 0.741708017808869 -0.09122188614116
## [16,] -0.342314121099202 0.78376539565179 0.512020607188603 -0.07872871062598
## [17,] 0.000006183686681 0.00001792069680 0.000020563556390 0.00000010802241
## [5,] [6,] [7,] [8,]
## [1,] -0.00092500311640 0.4652803404063 0.000476541187148 -0.0003074456185496
## [2,] -0.00544316693649 -0.7834992657177 -0.000437591208266 0.0004156446958949
## [3,] -0.00000001250728 -0.0000018003200 -0.000000001005486 0.0000000009550612
## [4,] 0.02692770685904 -0.4117496317030 0.000209517529386 0.0000428493239681
## [5,] 0.00000004177443 -0.0000001245475 -0.000025062140123 -0.0000050602271672
## [6,] -0.99921330114249 -0.0070690202978 0.003018729316630 -0.0002906679569835
## [7,] -0.00323299670836 0.0005212248736 -0.928208062774202 0.0930290019685839
## [8,] -0.00006246921838 -0.0001182740005 -0.371451710927692 -0.2612639639947784
## [9,] -0.00000083597251 0.0000009084183 -0.000539875515137 0.0000409608569991
## [10,] -0.00000082617401 0.0000002357646 -0.000783085527838 0.0000000857460508
## [11,] -0.00000224710303 0.0000012329735 -0.000464248881369 0.0003040838349096
## [12,] -0.00004169032543 0.0000640169553 -0.009076776876378 0.0224489384445046
## [13,] -0.00001629049033 -0.0004226816314 0.010628450812498 -0.9603369999865978
## [14,] 0.00000017565267 -0.0000003538937 0.000050240745243 -0.0000588135497904
## [15,] -0.02594990125873 -0.0049027311426 -0.000272758510803 0.0005541544397628
## [16,] -0.01154539312026 -0.0055557473969 0.002235045160704 0.0001513859595040
## [17,] 0.00020510603568 -0.0000605486210 0.015565937524174 -0.0182990788949705
## [9,] [10,] [11,]
## [1,] 0.0001605859903364 0.00002491936784812 0.00001601220645247
## [2,] -0.0001696089785111 -0.00002970998039459 -0.00003207715237611
## [3,] -0.0000000003897174 -0.00000000006829674 -0.00000000007396302
## [4,] 0.0000486769090081 -0.00002657789099886 -0.00000902005944176
## [5,] 0.0000054733425377 0.00027964584086260 0.00076541167174310
## [6,] 0.0011243911186266 0.00010271048128154 -0.00003856132743092
## [7,] -0.3584009689324924 0.03358636342809108 0.01336252181281341
## [8,] 0.8899264399971795 -0.03850425258139365 -0.01768203388937795
## [9,] -0.0002798272589561 -0.00132515557864803 -0.00791193281907092
## [10,] -0.0001600175428835 -0.00040427370362079 0.00204064618307306
## [11,] -0.0007107079906152 -0.00126841722462626 -0.00898079237252579
## [12,] -0.0248098862026032 -0.18747336017040794 -0.98157829401043206
## [13,] -0.2781666267929094 -0.00934492582032079 -0.01324433835270366
## [14,] 0.0001373777623987 0.00079790638674286 0.00203825629196005
## [15,] -0.0004387949233892 0.00002476411992577 -0.00005239734037122
## [16,] 0.0005902904363719 -0.00007526540226400 0.00001118365229877
## [17,] 0.0398118583643448 0.98089322969517778 -0.18890544010198079
## [12,] [13,] [14,] [15,]
## [1,] 0.00000002330388 0.00000002840513 -0.00000003785715 -0.00000007845501
## [2,] -0.00000026998355 0.00000055065828 0.00000018431264 0.00000025009820
## [3,] 0.00000073314919 -0.00000598048136 -0.00000753841920 0.00000251087786
## [4,] 0.00000021655590 -0.00000013826861 0.00000003037894 0.00000002783848
## [5,] 0.04173835214971 -0.01002457380511 -0.02528383441481 0.05748794849676
## [6,] 0.00000199247595 0.00000029588147 0.00000013085065 0.00000052940552
## [7,] -0.00073640461927 -0.00073874171442 -0.00007193205065 -0.00024158555124
## [8,] -0.00022095224077 0.00028154519593 0.00007289430839 -0.00011132696663
## [9,] 0.20730867888905 0.09676779395832 -0.04483561203595 0.97025166859262
## [10,] 0.95283982252208 0.20180149168765 0.02075959895647 -0.22436878008565
## [11,] -0.21763777294613 0.97312239856574 0.05370214442059 -0.04661422223617
## [12,] 0.00216721372513 -0.00912406113448 -0.00225772977126 -0.00786282170433
```

```
## [13,] -0.00003609868059 -0.00004708056327 -0.00001108996731 -0.00014139401553
## [14,] -0.00226846417160  0.05254051439851 -0.99701070142537 -0.05225500501467
## [15,]  0.00000025739419 -0.00000057809468 -0.00000009616703  0.00000007654168
## [16,]  0.00000094342987 -0.00000151677087  0.00000011618654 -0.00000052007679
## [17,]  0.00081169875269 -0.00027554945826  0.00040936332018 -0.00031608284324
##           [,16]           [,17]
## [1,]  0.00000002626414  0.00000000000000000000
## [2,] -0.00000008373433  0.00000229778966408858
## [3,]  0.00000235974679 -0.99999999994485777588
## [4,] -0.00000003810206  0.000000000000073452434
## [5,]  0.99710210942838  0.00000277840628666864
## [6,]  0.00000000286031  0.000000000000004087554
## [7,] -0.00000506471354  0.00000000380183176336
## [8,]  0.00002917137085 -0.00000000260595466882
## [9,] -0.06477535813131  0.00000219458929059154
## [10,] -0.02439577087212 -0.00000128572086477395
## [11,]  0.02295026131719 -0.00000644701776907673
## [12,]  0.00101972727436  0.00000005583887045888
## [13,]  0.00001861667946  0.000000000002761301082
## [14,] -0.02164735694476  0.00000701771608755439
## [15,] -0.00000001270250  0.000000000000453303872
## [16,]  0.00000007464302  0.00000000000775716831
## [17,] -0.00013815304227 -0.00000000196256816168
```

Then we do another certificate step to make sure the stability of the raw data and process data. If these two are very similar, then we can conclude that there is no bias between our dealing process.

```
s_dat <- scale(dat)
Mcor <- var(s_dat, na.rm = TRUE)
near(cor(dat), Mcor)
```

```
## SalePrice zhvi_px zhvi_idx AdjSalePrice NbrLivingUnits SqFtLot
## SalePrice TRUE TRUE TRUE TRUE TRUE TRUE
## zhvi_px TRUE TRUE TRUE TRUE TRUE TRUE
## zhvi_idx TRUE TRUE TRUE TRUE TRUE TRUE
## AdjSalePrice TRUE TRUE TRUE TRUE TRUE TRUE
## NbrLivingUnits TRUE TRUE TRUE TRUE TRUE TRUE
## SqFtLot TRUE TRUE TRUE TRUE TRUE TRUE
## SqFtTotLiving TRUE TRUE TRUE TRUE TRUE TRUE
## SqFtFinBasement TRUE TRUE TRUE TRUE TRUE TRUE
## Bathrooms TRUE TRUE TRUE TRUE TRUE TRUE
## Bedrooms TRUE TRUE TRUE TRUE TRUE TRUE
## BldgGrade TRUE TRUE TRUE TRUE TRUE TRUE
## YrBuilt TRUE TRUE TRUE TRUE TRUE TRUE
## YrRenovated TRUE TRUE TRUE TRUE TRUE TRUE
## TrafficNoise TRUE TRUE TRUE TRUE TRUE TRUE
## LandVal TRUE TRUE TRUE TRUE TRUE TRUE
## ImpsVal TRUE TRUE TRUE TRUE TRUE TRUE
## ZipCode TRUE TRUE TRUE TRUE TRUE TRUE
## SqFtTotLiving SqFtFinBasement Bathrooms Bedrooms BldgGrade
## SalePrice TRUE TRUE TRUE TRUE TRUE
## zhvi_px TRUE TRUE TRUE TRUE TRUE
## zhvi_idx TRUE TRUE TRUE TRUE TRUE
## AdjSalePrice TRUE TRUE TRUE TRUE TRUE
## NbrLivingUnits TRUE TRUE TRUE TRUE TRUE
## SqFtLot TRUE TRUE TRUE TRUE TRUE
## SqFtTotLiving TRUE TRUE TRUE TRUE TRUE
## SqFtFinBasement TRUE TRUE TRUE TRUE TRUE
## Bathrooms TRUE TRUE TRUE TRUE TRUE
## Bedrooms TRUE TRUE TRUE TRUE TRUE
## BldgGrade TRUE TRUE TRUE TRUE TRUE
## YrBuilt TRUE TRUE TRUE TRUE TRUE
## YrRenovated TRUE TRUE TRUE TRUE TRUE
## TrafficNoise TRUE TRUE TRUE TRUE TRUE
## LandVal TRUE TRUE TRUE TRUE TRUE
## ImpsVal TRUE TRUE TRUE TRUE TRUE
## ZipCode TRUE TRUE TRUE TRUE TRUE
## YrBuilt YrRenovated TrafficNoise LandVal ImpsVal ZipCode
## SalePrice TRUE TRUE TRUE TRUE TRUE TRUE
## zhvi_px TRUE TRUE TRUE TRUE TRUE TRUE
## zhvi_idx TRUE TRUE TRUE TRUE TRUE TRUE
## AdjSalePrice TRUE TRUE TRUE TRUE TRUE TRUE
## NbrLivingUnits TRUE TRUE TRUE TRUE TRUE TRUE
## SqFtLot TRUE TRUE TRUE TRUE TRUE TRUE
## SqFtTotLiving TRUE TRUE TRUE TRUE TRUE TRUE
## SqFtFinBasement TRUE TRUE TRUE TRUE TRUE TRUE
## Bathrooms TRUE TRUE TRUE TRUE TRUE TRUE
## Bedrooms TRUE TRUE TRUE TRUE TRUE TRUE
## BldgGrade TRUE TRUE TRUE TRUE TRUE TRUE
## YrBuilt TRUE TRUE TRUE TRUE TRUE TRUE
## YrRenovated TRUE TRUE TRUE TRUE TRUE TRUE
## TrafficNoise TRUE TRUE TRUE TRUE TRUE TRUE
## LandVal TRUE TRUE TRUE TRUE TRUE TRUE
## ImpsVal TRUE TRUE TRUE TRUE TRUE TRUE
## ZipCode TRUE TRUE TRUE TRUE TRUE TRUE
```

EigMcov and eigMcor are the results obtained after performing eigenvalue decomposition, corresponding to the covariance matrix Mcov and the correlation matrix Mcor, respectively.

```
eigMcor <- eigen(Mcor)
eigMcov
```

```
## eigen() decomposition
## $values
## [1] 326208688074.6403808593750 18235914108.6289024353027
## [3] 7213699330.8674087524414 2907594354.4751238822937
## [5] 819768275.0459656715393 265647813.2387284040451
## [7] 358827.5509719259571 193944.6326406990411
## [9] 139701.7964463515382 2380.4086192770487
## [11] 540.1920212516691 0.5116549668237
## [13] 0.3719691774605 0.2991274858044
## [15] 0.1831829428816 0.0231452300706
## [17] -0.0000002413329
##
## $vectors
##           [,1]           [,2]           [,3]           [,4]
## [1,] -0.602048190758632 -0.08632353123778 -0.348452357792817 -0.54053693604472
## [2,] -0.000564338517070 -0.00950159350215 -0.089000806375808 -0.61488790468656
## [3,] -0.000000001296734 -0.00000002183270 -0.000000204505529 -0.00000141288581
## [4,] -0.671371692965639 -0.08286364811084 -0.239609377807863 0.56099741735375
## [5,] -0.000000006144448 -0.00000003859377 -0.000000008546372 -0.00000002763985
## [6,] -0.006723293476451 0.00466043018045 -0.030830105151754 0.02224672094794
## [7,] -0.001135948383779 0.00202124256211 0.000533277664980 0.00007708117987
## [8,] -0.000234642301026 0.00000292508448 0.000491919926047 -0.00007472283429
## [9,] -0.000000729351419 0.00000185276710 0.000000243518647 0.00000013819191
## [10,] -0.000000506493967 0.00000099188393 0.000000216660954 -0.00000014412608
## [11,] -0.000001427115810 0.00000261541785 0.000001065153434 0.00000047833112
## [12,] -0.000004410954214 0.00006509561867 -0.000045394841596 0.00001309590428
## [13,] -0.000088015559774 -0.00001345352307 0.000496762583082 -0.00011173299429
## [14,] 0.000000010813782 -0.00000005661127 0.000000127913354 -0.00000009564167
## [15,] -0.263765414693926 -0.60932551094283 0.741708017808869 -0.09122188614116
## [16,] -0.342314121099202 0.78376539565179 0.512020607188603 -0.07872871062598
## [17,] 0.000006183686681 0.00001792069680 0.000020563556390 0.00000010802241
##           [,5]           [,6]           [,7]           [,8]
## [1,] -0.00092500311640 0.4652803404063 0.000476541187148 -0.0003074456185496
## [2,] -0.00544316693649 -0.7834992657177 -0.000437591208266 0.0004156446958949
## [3,] -0.00000001250728 -0.0000018003200 -0.000000001005486 0.0000000009550612
## [4,] 0.02692770685904 -0.4117496317030 0.000209517529386 0.0000428493239681
## [5,] 0.00000004177443 -0.0000001245475 -0.000025062140123 -0.0000050602271672
## [6,] -0.99921330114249 -0.0070690202978 0.003018729316630 -0.0002906679569835
## [7,] -0.00323299670836 0.0005212248736 -0.928208062774202 0.0930290019685839
## [8,] -0.00006246921838 -0.0001182740005 -0.371451710927692 -0.2612639639947784
## [9,] -0.00000083597251 0.0000009084183 -0.000539875515137 0.0000409608569991
## [10,] -0.00000082617401 0.0000002357646 -0.000783085527838 0.0000000857460508
## [11,] -0.00000224710303 0.0000012329735 -0.000464248881369 0.0003040838349096
## [12,] -0.00004169032543 0.0000640169553 -0.009076776876378 0.0224489384445046
## [13,] -0.00001629049033 -0.0004226816314 0.010628450812498 -0.9603369999865978
## [14,] 0.00000017565267 -0.0000003538937 0.000050240745243 -0.0000588135497904
## [15,] -0.02594990125873 -0.0049027311426 -0.000272758510803 0.0005541544397628
## [16,] -0.01154539312026 -0.0055557473969 0.002235045160704 0.0001513859595040
## [17,] 0.00020510603568 -0.0000605486210 0.015565937524174 -0.0182990788949705
##           [,9]           [,10]           [,11]
## [1,] 0.0001605859903364 0.00002491936784812 0.00001601220645247
## [2,] -0.0001696089785111 -0.00002970998039459 -0.00003207715237611
## [3,] -0.0000000003897174 -0.00000000006829674 -0.00000000007396302
## [4,] 0.0000486769090081 -0.00002657789099886 -0.00000902005944176
## [5,] 0.0000054733425377 0.00027964584086260 0.00076541167174310
## [6,] 0.0011243911186266 0.00010271048128154 -0.00003856132743092
## [7,] -0.3584009689324924 0.03358636342809108 0.01336252181281341
## [8,] 0.8899264399971795 -0.03850425258139365 -0.01768203388937795
## [9,] -0.0002798272589561 -0.00132515557864803 -0.00791193281907092
## [10,] -0.0001600175428835 -0.00040427370362079 0.00204064618307306
## [11,] -0.0007107079906152 -0.00126841722462626 -0.00898079237252579
## [12,] -0.0248098862026032 -0.18747336017040794 -0.98157829401043206
## [13,] -0.2781666267929094 -0.00934492582032079 -0.01324433835270366
## [14,] 0.0001373777623987 0.00079790638674286 0.00203825629196005
## [15,] -0.0004387949233892 0.00002476411992577 -0.00005239734037122
## [16,] 0.0005902904363719 -0.00007526540226400 0.00001118365229877
## [17,] 0.0398118583643448 0.98089322969517778 -0.18890544010198079
##           [,12]           [,13]           [,14]           [,15]
## [1,] 0.00000002330388 0.00000002840513 -0.00000003785715 -0.00000007845501
## [2,] -0.00000026998355 0.00000055065828 0.00000018431264 0.00000025009820
## [3,] 0.00000073314919 -0.00000598048136 -0.00000753841920 0.00000251087786
## [4,] 0.00000021655590 -0.00000013826861 0.00000003037894 0.00000002783848
## [5,] 0.04173835214971 -0.01002457380511 -0.02528383441481 0.05748794849676
## [6,] 0.00000199247595 0.00000029588147 0.00000013085065 0.00000052940552
## [7,] -0.00073640461927 -0.00073874171442 -0.00007193205065 -0.00024158555124
## [8,] -0.00022095224077 0.00028154519593 0.00007289430839 -0.00011132696663
## [9,] 0.20730867888905 0.09676779395832 -0.04483561203595 0.97025166859262
## [10,] 0.95283982252208 0.20180149168765 0.02075959895647 -0.22436878008565
## [11,] -0.21763777294613 0.97312239856574 0.05370214442059 -0.04661422223617
## [12,] 0.00216721372513 -0.00912406113448 -0.00225772977126 -0.00786282170433
```



```
## [13,] -0.00003609868059 -0.00004708056327 -0.00001108996731 -0.00014139401553
## [14,] -0.00226846417160  0.05254051439851 -0.99701070142537 -0.05225500501467
## [15,]  0.00000025739419 -0.00000057809468 -0.00000009616703  0.00000007654168
## [16,]  0.00000094342987 -0.00000151677087  0.00000011618654 -0.00000052007679
## [17,]  0.00081169875269 -0.00027554945826  0.00040936332018 -0.00031608284324
##           [,16]           [,17]
## [1,]  0.00000002626414  0.00000000000000000000
## [2,] -0.00000008373433  0.00000229778966408858
## [3,]  0.00000235974679 -0.99999999994485777588
## [4,] -0.00000003810206  0.000000000000073452434
## [5,]  0.99710210942838  0.00000277840628666864
## [6,]  0.00000000286031  0.000000000000004087554
## [7,] -0.00000506471354  0.00000000380183176336
## [8,]  0.00002917137085 -0.00000000260595466882
## [9,] -0.06477535813131  0.00000219458929059154
## [10,] -0.02439577087212 -0.00000128572086477395
## [11,]  0.02295026131719 -0.00000644701776907673
## [12,]  0.00101972727436  0.00000005583887045888
## [13,]  0.00001861667946  0.000000000002761301082
## [14,] -0.02164735694476  0.00000701771608755439
## [15,] -0.00000001270250  0.000000000000453303872
## [16,]  0.00000007464302  0.00000000000775716831
## [17,] -0.00013815304227 -0.00000000196256816168
```

eigMcor

```

## eigen() decomposition
## $values
## [1] 5.6018600693110149890686 2.0298722169467926512709 1.9191278147356185534989
## [4] 1.3001771610606518425612 1.0007811293439694555474 0.9948462230660555238870
## [7] 0.8986962933955856547286 0.8947928084106385737684 0.7771801729175958906026
## [10] 0.5203899109238445674919 0.3377168188234225820743 0.2545110391917222991687
## [13] 0.2232965768879534851354 0.1301758177684894657311 0.1120751402842977556551
## [16] 0.0045008069321234907972 0.0000000000000004451719
##
## $vectors
##           [,1]      [,2]      [,3]      [,4]      [,5]      [,6]
## [1,] -0.37819612  0.13941611  0.12456037 -0.228851824  0.09289270 -0.012398619
## [2,]  0.01014419  0.65682061 -0.25287326  0.002279683 -0.01328419  0.003540483
## [3,]  0.01014419  0.65682061 -0.25287326  0.002279683 -0.01328419  0.003540483
## [4,] -0.38092191  0.04728309  0.16287703 -0.229931578  0.09498655 -0.013345328
## [5,] -0.02174748  0.05936637  0.10301880  0.517283126  0.32963208 -0.155170320
## [6,] -0.07938282 -0.03836911 -0.07885330 -0.034137136 -0.22476482 -0.870968346
## [7,] -0.37997752 -0.03382650 -0.08412441  0.177644281 -0.06523639  0.003972539
## [8,] -0.16308421  0.10008297  0.28726574  0.365129387 -0.19697382  0.079455965
## [9,] -0.32655611 -0.08159000 -0.19417930  0.261320820 -0.02163761  0.089528459
## [10,] -0.22902257 -0.02217832 -0.09000857  0.518929762 -0.13802451  0.050139929
## [11,] -0.35243459 -0.08052204 -0.15345851 -0.096070162  0.05876722  0.050793366
## [12,] -0.11913813 -0.20989614 -0.55401151 -0.029674582  0.11848495  0.093552656
## [13,] -0.04603084  0.09465634  0.27324059  0.050292090 -0.56560876 -0.093136061
## [14,]  0.02204315  0.07711263  0.19242687  0.134623207  0.63799612 -0.322880153
## [15,] -0.29659364  0.09562150  0.26191903 -0.276892786  0.07606004 -0.042102626
## [16,] -0.37391093  0.01393657  0.06494528 -0.107409838  0.06270892  0.075146715
## [17,]  0.07765466  0.13194914  0.40353856  0.067659928  0.07799660  0.261616866
##           [,7]      [,8]      [,9]      [,10]     [,11]
## [1,]  0.11498710  0.005427198 -0.02473347  0.027903435 -0.079249906
## [2,] -0.01986510  0.005500685  0.01533887 -0.006182737  0.005665891
## [3,] -0.01986510  0.005500685  0.01533887 -0.006182737  0.005665891
## [4,]  0.11850310  0.004427628 -0.02613710  0.029502274 -0.078364617
## [5,]  0.65598015 -0.200945770  0.21141837 -0.227808805  0.052117951
## [6,] -0.01586907  0.320435556  0.24848200  0.036137503 -0.069023266
## [7,] -0.10043582  0.056945568  0.01475014  0.000849951  0.245786849
## [8,] -0.22907385  0.419445952 -0.30993603 -0.555549357 -0.045327651
## [9,] -0.11209166 -0.090829304  0.15174944 -0.141507917 -0.229039426
## [10,] -0.02500777  0.033159519 -0.20807072  0.706195099 -0.143919784
## [11,] -0.10582503 -0.067336687  0.19934242 -0.016279291  0.338313298
## [12,] -0.11905987 -0.071580116  0.17840893 -0.265933988 -0.508949857
## [13,] -0.14803522 -0.703142601  0.15019220 -0.107460570 -0.108783540
## [14,] -0.57060399 -0.281040599 -0.13977768  0.026968800 -0.040713508
## [15,]  0.24166176 -0.010198751 -0.26884447  0.056349186 -0.474681402
## [16,] -0.06330338 -0.023070086  0.19209903 -0.013745084  0.407549039
## [17,] -0.17420269  0.290890935  0.71158432  0.176280522 -0.256904774
##           [,12]     [,13]     [,14]     [,15]     [,16]
## [1,]  0.104675320 -0.2460435550  0.003183568 -0.4137202039  0.7051725594
## [2,] -0.009988429  0.0209079808  0.004828592  0.0236768063 -0.0499338203
## [3,] -0.009988429  0.0209079808  0.004828592  0.0236768063 -0.0499338203
## [4,]  0.105292018 -0.2479946234  0.002555553 -0.4147035811 -0.7055028973
## [5,] -0.109091291 -0.0547072122  0.007601236  0.0109988910  0.0005188216
## [6,]  0.013918051 -0.0008035467  0.059269846  0.0358659885  0.0002341412
## [7,] -0.055871663  0.1067125236 -0.849449746  0.0391533519  0.0004517259
## [8,] -0.162828987 -0.1166546522  0.163719696 -0.0186650433 -0.0002314837
## [9,]  0.622914644  0.4780554325  0.184385823 -0.0730907954 -0.0006969019
## [10,] -0.129598350 -0.1869077876  0.177998363  0.0156794345 -0.0002465395
## [11,] -0.593428012  0.3993386340  0.342016274 -0.1780394994  0.0008380123
## [12,] -0.254591352 -0.3871896626 -0.060373911  0.1090401086 -0.0016427036
## [13,] -0.104864165 -0.0966451283 -0.021036602 -0.0100636592  0.0004070034
## [14,] -0.013144158 -0.0074923718 -0.020012310 -0.0008543371  0.0001435488
## [15,] -0.192931956  0.3562773383 -0.052981143  0.4730905271  0.0015549659
## [16,]  0.239210629 -0.3730105123  0.226325107  0.6159147257  0.0016991545
## [17,] -0.092854255  0.0277858868 -0.082596752  0.0017651955  0.0007282313
##           [,17]
## [1,] -0.00000000000027410768339
## [2,]  0.70710678118655623247690
## [3,] -0.70710678118653857993081
## [4,]  0.00000000000026766244976
## [5,]  0.00000000000000042411278
## [6,]  0.000000000000000033711381
## [7,]  0.00000000000000038458335
## [8,]  0.00000000000000016009621
## [9,]  0.00000000000000032825389
## [10,] -0.00000000000000028914300
## [11,]  0.00000000000000007075445
## [12,]  0.000000000000000123258007
## [13,] -0.00000000000000018174601
## [14,] -0.00000000000000052479437
## [15,]  0.000000000000000327976823

```

```
## [16,] 0.000000000000000198348722  
## [17,] 0.00000000000000012802749
```

Then we can calculate the loading matrix to find the relationship between origin variable and PCs. Use Sigma to confirm we need six PCs, and use loading^2 to check which variable is important in each PC.

By using loading^2 , we can decide to keep what variables in PCs.

```
loadings <- t(t(eigMcor$vector)*sqrt(eigMcor$values))  
row.names(loadings)<-colnames(dat)  
colnames(loadings)<-paste0("PC",seq(1:ncol(loadings)))  
loadings <- as.data.frame(loadings)  
loadings ^ 2
```

##	PC1	PC2	PC3	PC4
## SalePrice	0.8012469783	0.0394543236	0.02977582	0.068094382933
## zhvi_px	0.0005764576	0.8757138976	0.12271841	0.000006756959
## zhvi_idx	0.0005764576	0.8757138976	0.12271841	0.000006756959
## AdjSalePrice	0.8128383038	0.0045381662	0.05091240	0.068738456243
## NbrLivingUnits	0.0026494171	0.0071540116	0.02036746	0.347903786591
## SqFtLot	0.0353008645	0.0029883540	0.01193284	0.001515153749
## SqFtTotLiving	0.8088128713	0.0023226446	0.01358151	0.041030328561
## SqFtFinBasement	0.1489896533	0.0203324179	0.15836951	0.173338929462
## Bathrooms	0.5973761603	0.0135127135	0.07236186	0.088787240364
## Bedrooms	0.2938250534	0.0009984496	0.01554790	0.350122234749
## BldgGrade	0.6958078330	0.0131612822	0.04519453	0.011999953982
## YrBuilt	0.0795122125	0.0894288381	0.58903551	0.001144911066
## YrRenovated	0.0118694345	0.0181872949	0.14328289	0.003288530729
## TrafficNoise	0.0027219459	0.0120703460	0.07106166	0.023563641006
## LandVal	0.4927832397	0.0185600793	0.13165520	0.099684081961
## ImpsVal	0.7831925870	0.0003942581	0.00809467	0.014999979131
## ZipCode	0.0337805995	0.0353412423	0.31251724	0.005952036615
##	PC5	PC6	PC7	PC8
## SalePrice	0.0086357950	0.00015293349	0.0118825930	0.00002635565
## zhvi_px	0.0001766076	0.00001247041	0.0003546457	0.00002707422
## zhvi_idx	0.0001766076	0.00001247041	0.0003546457	0.00002707422
## AdjSalePrice	0.0090294932	0.00017717990	0.0126203790	0.00001754142
## NbrLivingUnits	0.1087421861	0.02395373650	0.3867179663	0.03613101994
## SqFtLot	0.0505586872	0.75467627771	0.0002263165	0.09187638192
## SqFtTotLiving	0.0042591106	0.00001569973	0.0090654673	0.00290163211
## SqFtFinBasement	0.0388289918	0.00628071329	0.0471589359	0.15742528907
## Bathrooms	0.0004685517	0.00797403573	0.0112917052	0.00738200710
## Bedrooms	0.0190656462	0.00250105584	0.0005620346	0.00098387276
## BldgGrade	0.0034562838	0.00256666947	0.0100644433	0.00405719586
## YrBuilt	0.0140496499	0.00870699303	0.0127392469	0.00458466160
## YrRenovated	0.3201631676	0.00862962032	0.0196944127	0.44239408109
## TrafficNoise	0.4073569997	0.10371430380	0.2926055529	0.07067415259
## LandVal	0.0057896488	0.00176349535	0.0524842306	0.00009307144
## ImpsVal	0.0039354804	0.00561792531	0.0036013617	0.00047623457
## ZipCode	0.0060882222	0.06809064277	0.0272723562	0.07571516284
##	PC9	PC10	PC11	PC12
## SalePrice	0.0004754357	0.0004051764641	0.00212104655	0.00278865777
## zhvi_px	0.0001828557	0.0000198925493	0.00001084149	0.00002539224
## zhvi_idx	0.0001828557	0.0000198925493	0.00001084149	0.00002539224
## AdjSalePrice	0.0005309290	0.0004529391518	0.00207392345	0.00282161348
## NbrLivingUnits	0.0347381886	0.0270065980411	0.00091733372	0.00302891289
## SqFtLot	0.0479856737	0.0006795871268	0.00160895426	0.00004930188
## SqFtTotLiving	0.0001690886	0.0000003759384	0.02040186986	0.00079449252
## SqFtFinBasement	0.0746561944	0.1606105857511	0.00069387159	0.00674792221
## Bathrooms	0.0178968225	0.0104205429135	0.01771630636	0.09875604867
## Bedrooms	0.0336467916	0.2595244423645	0.00699509611	0.00427469932
## BldgGrade	0.0308831191	0.0001379112922	0.03865367835	0.08962779461
## YrBuilt	0.0247374479	0.0368024355504	0.08747877305	0.01649658004
## YrRenovated	0.0175313938	0.0060093451534	0.00399649307	0.00279872889
## TrafficNoise	0.0151843915	0.0003784879882	0.00055979593	0.00004397159
## LandVal	0.0561725230	0.0016523580825	0.07609517541	0.00947359816
## ImpsVal	0.0286795332	0.0000983158761	0.05609348684	0.01456356075
## ZipCode	0.3935269288	0.0161710241312	0.02228933128	0.00219437193
##	PC13	PC14	PC15	
## SalePrice	0.0135178011039	0.0000013193455	0.01918327493805	
## zhvi_px	0.0000976126830	0.0000030350881	0.00006282833268	
## zhvi_idx	0.0000976126830	0.0000030350881	0.00006282833268	
## AdjSalePrice	0.0137330371907	0.0000008501587	0.01927457729169	
## NbrLivingUnits	0.0006682996503	0.0000075214008	0.00001355835766	
## SqFtLot	0.0000001441798	0.0004572965300	0.00014417000114	
## SqFtTotLiving	0.0025428037699	0.0939302971727	0.00017180950461	
## SqFtFinBasement	0.0030386885643	0.0034892506769	0.00003904516778	
## Bathrooms	0.0510315290205	0.0044257346107	0.00059873502956	
## Bedrooms	0.0078007589720	0.0041244147532	0.00002755307536	
## BldgGrade	0.0356094053702	0.0152273334500	0.00355256489449	
## YrBuilt	0.0334756927436	0.0004744920427	0.00133254487098	
## YrRenovated	0.0020856527354	0.0000576078297	0.00001135066040	
## TrafficNoise	0.0000125348951	0.0000521344468	0.00000008180274	
## LandVal	0.0283438253714	0.0003654037231	0.02508405793950	
## ImpsVal	0.0310687806026	0.0066680029375	0.04251581086832	
## ZipCode	0.0001723973521	0.0008880885140	0.00000034921665	
##	PC16			
## SalePrice	0.0022381087853359			
## zhvi_px	0.0000112222508495			
## zhvi_idx	0.0000112222508497			
## AdjSalePrice	0.0022402061595203			
## NbrLivingUnits	0.0000000012115086			
## SqFtLot	0.000000002467436			
## SqFtTotLiving	0.0000000009184182			

[illegible]

```
(sigma <- round(eigMcor$values/ncol(dat),2))
```

```
## [1] 0.33 0.12 0.11 0.08 0.06 0.06 0.05 0.05 0.05 0.03 0.02 0.01 0.01 0.01 0.01
## [16] 0.00 0.00
```

```
(sigmaSum <- cumsum(sigma))
```

```
## [1] 0.33 0.45 0.56 0.64 0.70 0.76 0.81 0.86 0.91 0.94 0.96 0.97 0.98 0.99 1.00
## [16] 1.00 1.00
```

So we use these 9 variables to do the second PCA.

PC1: Sale Price, Adj Sale Price, SqFtTotLiving, ImpsVal

PC2: Bedrooms

PC3: TrafficNoise

PC4: YrRenovated

PC5: SqFtLot

PC6: NbrLibingUnits

```
selected_variables <- list(
  PC1 = c("SalePrice", "AdjSalePrice", "SqFtTotLiving", "ImpsVal"),
  PC2 = c("Bedrooms"),
  PC3 = c("TrafficNoise"),
  PC4 = c("YrRenovated"),
  PC5 = c("SqFtLot"),
  PC6 = c("NbrLivingUnits")
)
selected_data <- dat[, unlist(selected_variables)]
pca_mod_second <- pca(selected_data, nfactors = ncol(selected_data), covar = FALSE, rotate = "none")
pca_mod_second$loadings
```

```
##
## Loadings:
##      PC1      PC2      PC3      PC4      PC5      PC6      PC7      PC8      PC9
## SalePrice    0.926 -0.126  0.142          -0.125 -0.190  0.207
## AdjSalePrice  0.930 -0.127  0.142          -0.122 -0.189  0.195
## SqFtTotLiving 0.879          -0.171          0.194 -0.313 -0.236
## ImpsVal       0.908          0.105          -0.232  0.293
## Bedrooms     0.543  0.418 -0.371 -0.178  0.138  0.542  0.203
## TrafficNoise  0.437  0.623  0.468 -0.334  0.298
## YrRenovated   0.145          0.584 -0.236  0.761
## SqFtLot       0.209          -0.284  0.813  0.449
## NbrLivingUnits      0.839 -0.129          -0.517
##
##      PC1      PC2      PC3      PC4      PC5      PC6      PC7      PC8      PC9
## SS loadings  3.685  1.126  1.044  0.972  0.955  0.769  0.276  0.158  0.014
## Proportion Var 0.409  0.125  0.116  0.108  0.106  0.085  0.031  0.018  0.002
## Cumulative Var 0.409  0.535  0.651  0.759  0.865  0.950  0.981  0.998  1.000
```

So we can get a new PCA results, which can suggest the variables in each PC more clearly. And the cumulative sigma value suggests that we should choose PC1 - PC5 as our results.

PC1: SalePrice, AdjSalePrice, SqFtTotLiving, ImpsVal

PC2: NbrLibingUnits

PC3: TrafficNoise

PC4: SqFtLot

PC5: YrRenovated

So with the second PCA, we finally made 6 PC into 5 PC, and make each PC’s variable become more clear.