

Craigslist apartments

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Overview

We wish to build an application that helps a user find out the price and size of the apartment that the user is likely to get, given a city and the desired number of bedrooms.

To do this we are going to obtain recent data from craigslist's "apt/housing" for a few United States cities and use it to train a model to predict price and size.

Data processing

It may not be exactly enough data, but lets try doing this with just the 100 most recent posts from each city. `get_craigslist_data.R` was used to download the data, and it stored into `data/citiesdf.rds`

Print summary.

```
citiesdf <- readRDS('data/citiesdf.rds')
str(citiesdf)
```

```
## 'data.frame':   800 obs. of  7 variables:
## $ city      : Factor w/ 8 levels "chicago","honolulu",...: 1 1 1 1 1 1 1 1 1 ...
## $ date      : Factor w/ 1 level "Nov 21": 1 1 1 1 1 1 1 1 1 ...
## $ title     : Factor w/ 740 levels "**SHORT TERM!** 1 bedroom Located in Lincoln Park",...: 26 62 1 49
## $ price     : num  2848 985 1295 1225 1300 ...
## $ bedrooms: num  2 NA 1 3 2 NA 1 4 1 2 ...
## $ sqft      : num  1123 NA NA NA NA ...
## $ href      : Factor w/ 800 levels "http://chicago.craigslist.org/chc/apa/5283655742.html",...: 76 62 0
```

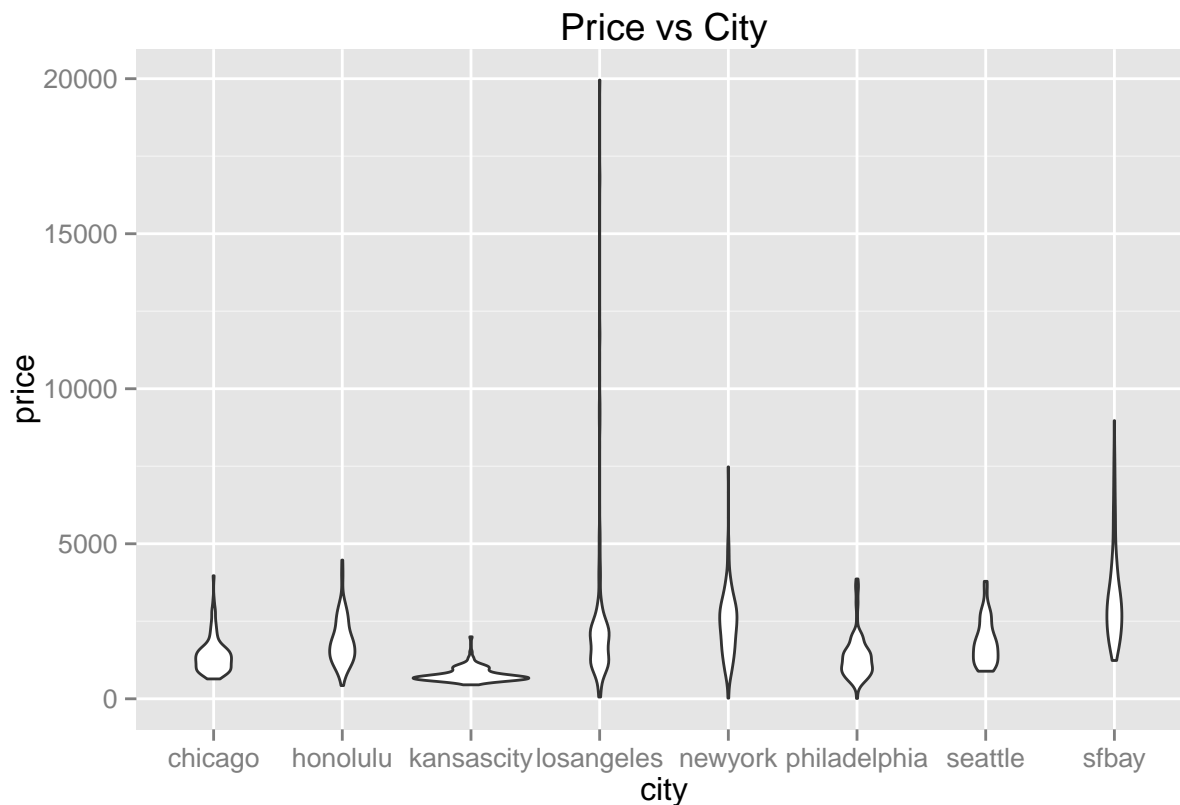
```
summary(citiesdf)
```

```
##           city           date
##  chicago      :100   Nov 21:800
##  honolulu     :100
##  kansascity   :100
##  losangeles   :100
##  newyork      :100
##  philadelphia:100
##  (Other)      :200
##
##                                     title
##  The Park Apartments                : 17
##  Your New Home Is Waiting in Olathe, KS      : 8
##  Live in the Heart of It All, Community Clubhouse, Quartz Countertops: 7
##  Beverly Plaza Apartments                : 6
##  1BR LEFFERT GARDEN NEAR EVERYTHING        : 5
##  Centrally Located Studio, 1 Bath in Lakeview, Available: Now      : 3
```

```
## (Other) :754
## price bedrooms sqft
## Min. : 1 Min. :1.000 Min. : 175
## 1st Qu.: 1032 1st Qu.:1.000 1st Qu.: 700
## Median : 1595 Median :2.000 Median : 904
## Mean : 1928 Mean :1.954 Mean :1038
## 3rd Qu.: 2406 3rd Qu.:2.000 3rd Qu.:1200
## Max. :20010 Max. :8.000 Max. :4490
## NA's :101 NA's :300
## href
## http://chicago.craigslist.org/chc/apa/5283655742.html: 1
## http://chicago.craigslist.org/chc/apa/5285745015.html: 1
## http://chicago.craigslist.org/chc/apa/5286189411.html: 1
## http://chicago.craigslist.org/chc/apa/5286848681.html: 1
## http://chicago.craigslist.org/chc/apa/5286911383.html: 1
## http://chicago.craigslist.org/chc/apa/5286912144.html: 1
## (Other) :794
```

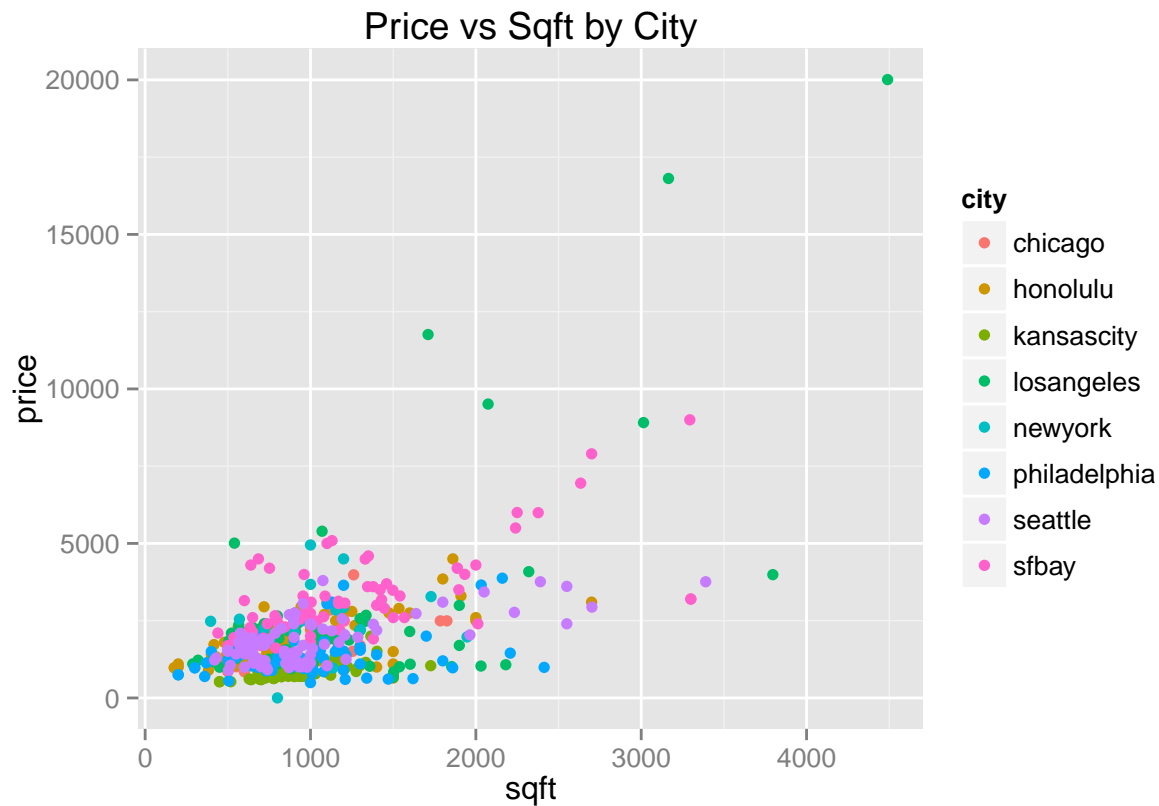
Exploratory plots

```
library(ggplot2)
ggplot(citiesdf, aes(y=price, x=city)) + geom_violin() +
  labs(title="Price vs City")
```



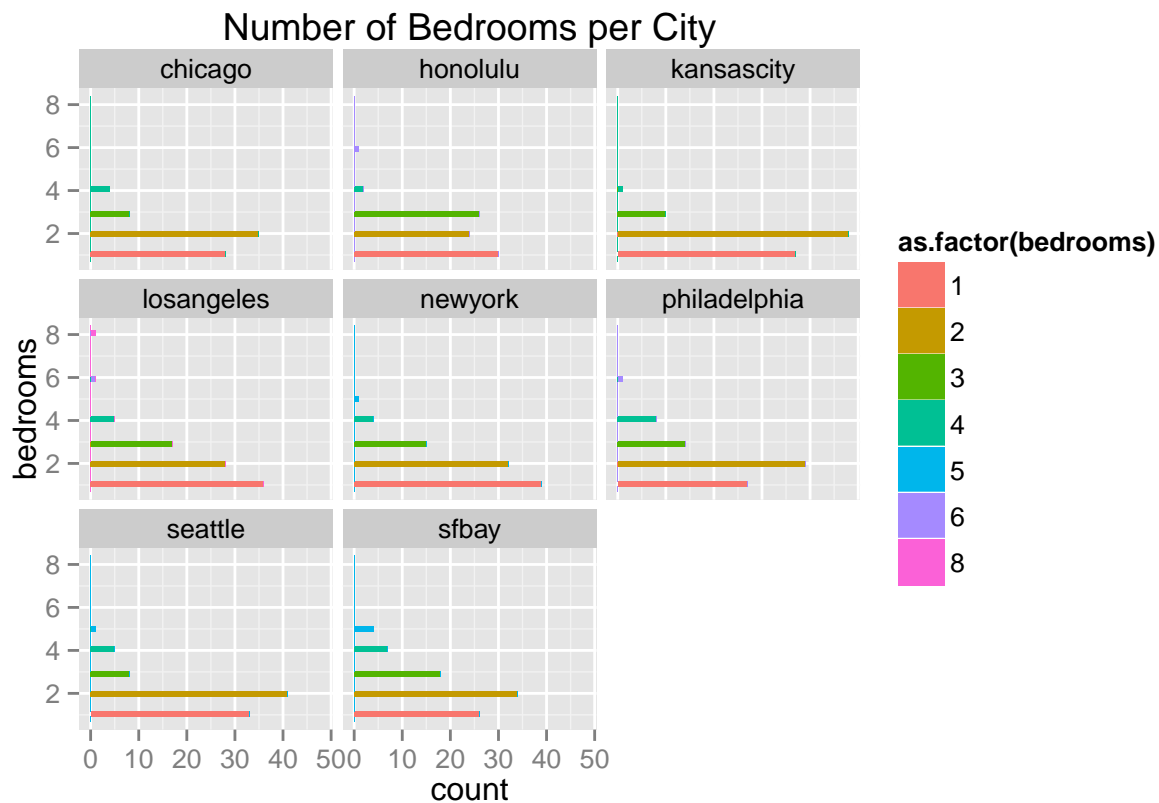
```
ggplot(citiesdf, aes(x=sqft, y=price, color=city)) + geom_point() +
  labs(title="Price vs Sqft by City")
```

Warning: Removed 300 rows containing missing values (geom_point).



```
ggplot(citiesdf, aes(x=bedrooms, fill=as.factor(bedrooms))) + geom_histogram() +
  coord_flip() + facet_wrap(~ city) +
  labs(title="Number of Bedrooms per City")
```

stat_bin: binwidth defaulted to range/30. Use 'binwidth = x' to adjust this.
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Model training

train_models.R was used to train two models, which are as follows.

model formula	saved to file
price ~ city + bedrooms	data/pricemod.rds
sqft ~ city + bedrooms	data/sqftmod.rds

```
pricemod <- readRDS('data/pricemod.rds')
summary(pricemod)
```

```
##
## Call:
## lm(formula = price ~ city + bedrooms, data = citiestdf)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2575.5  -523.2  -117.4   274.1 16421.5
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      639.51     174.18   3.672 0.000260 ***
```

```
## cityhonolulu      297.52      203.51      1.462 0.144208
## citykansascity    -730.79      196.66     -3.716 0.000219 ***
## citylosangeles    955.74      200.67      4.763 2.33e-06 ***
## citynewyork       893.34      198.96      4.490 8.34e-06 ***
## cityphiladelphia -272.08      200.32     -1.358 0.174840
## cityseattle       321.24      200.48      1.602 0.109543
## citysfbay         1691.29      200.80      8.423 < 2e-16 ***
## bedrooms          498.32       50.52      9.864 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1276 on 690 degrees of freedom
## (101 observations deleted due to missingness)
## Multiple R-squared:  0.3414, Adjusted R-squared:  0.3337
## F-statistic: 44.7 on 8 and 690 DF, p-value: < 2.2e-16
```

```
sqftmod <- readRDS('data/sqftmod.rds')
summary(sqftmod)
```

```
##
## Call:
## lm(formula = sqft ~ city + bedrooms, data = citiesdf)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -1585.32  -168.28   -17.71    98.83   2402.07
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      207.29      77.66   2.669  0.00788 **
## cityhonolulu     -157.92      81.81  -1.930  0.05422 .
## citykansascity    -31.11      81.90  -0.380  0.70420
## citylosangeles     63.25      81.79   0.773  0.43975
## citynewyork       -14.82     102.97  -0.144  0.88565
## cityphiladelphia   20.93      82.41   0.254  0.79964
## cityseattle       -24.93      79.35  -0.314  0.75355
## citysfbay         31.60      80.85   0.391  0.69609
## bedrooms         454.35      15.60  29.117 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 313 on 440 degrees of freedom
## (351 observations deleted due to missingness)
## Multiple R-squared:  0.6791, Adjusted R-squared:  0.6732
## F-statistic: 116.4 on 8 and 440 DF, p-value: < 2.2e-16
```

Check out some predictions

```
test <- data.frame(
  city = rep(c('chicago', 'honolulu', 'kansascity', 'losangeles', 'newyork', 'philadelphia', 'sfbay'),
    bedrooms = rep(c(1,2,3)))
```

```
test <- test[order(test$city, test$bedrooms),]

# Predict price (USD) and size (sqft).
cbind(test,
      predictedPrice=predict(pricemod, newdata = test),
      predictedSize=predict(sqftmod, newdata = test))
```

##	city	bedrooms	predictedPrice	predictedSize
## 1	chicago	1	1137.8344	661.6365
## 17	chicago	2	1636.1585	1115.9839
## 9	chicago	3	2134.4827	1570.3314
## 10	honolulu	1	1435.3590	503.7144
## 2	honolulu	2	1933.6832	958.0618
## 18	honolulu	3	2432.0073	1412.4093
## 19	kansascity	1	407.0415	630.5227
## 11	kansascity	2	905.3657	1084.8701
## 3	kansascity	3	1403.6898	1539.2175
## 4	losangeles	1	2093.5699	724.8854
## 20	losangeles	2	2591.8940	1179.2329
## 12	losangeles	3	3090.2182	1633.5803
## 13	newyork	1	2031.1727	646.8192
## 5	newyork	2	2529.4969	1101.1667
## 21	newyork	3	3027.8210	1555.5141
## 22	philadelphia	1	865.7515	682.5663
## 14	philadelphia	2	1364.0756	1136.9137
## 6	philadelphia	3	1862.3998	1591.2611
## 16	seattle	1	1459.0723	636.7086
## 8	seattle	2	1957.3965	1091.0561
## 24	seattle	3	2455.7206	1545.4035
## 7	sfbay	1	2829.1272	693.2376
## 23	sfbay	2	3327.4513	1147.5850
## 15	sfbay	3	3825.7754	1601.9324