Lab₁

AUTHOR

JP Garcia

Question 1: Part a

```
library(readr)
rent_data <- read_csv("~/jp-garcia-131a/craigslist.csv")

Rows: 5876 Columns: 7
   — Column specification
Delimiter: ","
chr (3): title, link, location
dbl (3): price, size, brs
dttm (1): time

i Use `spec()` to retrieve the full column specification for this data.
i Specify the column types or set `show_col_types = FALSE` to quiet this message.

num_postings <- nrow(rent_data)
print(paste("The number of postings in the craiglist dataset is ", num_postings))</pre>
```

[1] "The number of postings in the craiglist dataset is 5876"

Question 1: Part b

```
#calculation
mean_rent <- mean(rent_data$price)
median_rent <- median(rent_data$price)
max_rent <- max(rent_data$price)
min_rent <- min(rent_data$price)

print(paste("Mean Monthly Rent:", mean_rent))</pre>
```

[1] "Mean Monthly Rent: 3125.22515316542"

```
print(paste("Median Monthly Rent:", median_rent))
```

[1] "Median Monthly Rent: 2865"

```
print(paste("Max Monthly Rent:", max_rent))
```

[1] "Max Monthly Rent: 20000"

```
print(paste("Min Monthly Rent :", min_rent))
```

```
[1] "Min Monthly Rent: 600"
```

Question 1: Part c

```
city_postings <- table(rent_data$location)
print("City Distribution:")</pre>
```

[1] "City Distribution:"

```
print(city_postings)
```

alameda	albany / el cerrito	berkeley	emeryville
200	111	396	210
menlo park	mountain view	oakland	palo alto
389	1052	881	710
${\sf redwood}\ {\sf city}$	richmond	sunnyvale	
521	451	955	

Question 1: Part d

```
total_postings <- nrow(rent_data)
postings_over_3000 <- sum(rent_data$price > 3000)
percent_over_3000 <- (postings_over_3000 / total_postings) * 100

print(paste("Entries Over 3000:", round(percent_over_3000, 2), "%"))</pre>
```

[1] "Entries Over 3000: 42.31 %"

Question 2

```
library(ggplot2)
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

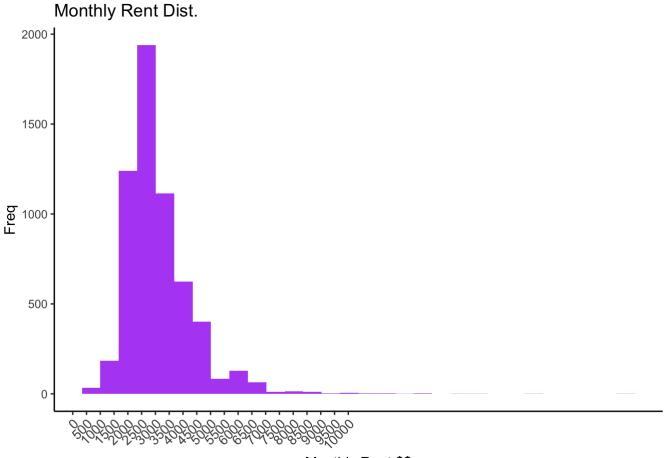
```
craigslist <- read.csv("~/jp-garcia-131a/craigslist.csv")
craigslist %>%
```

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```
ggplot(aes(x = price)) +
geom_histogram(bins = 30, fill = "purple", alpha = 0.8) +
lims(x = c(0, 10000)) +
scale_x_continuous(breaks = seq(from = 0, to = 10000, by = 500)) +
labs(
    title = "Monthly Rent Dist.",
    x = "Monthly Rent $$",
    y = "Freq"
) +
theme_classic() +
theme(axis.text.x = element_text(angle = 45, hjust = 1, size = 10))
```

Lab 1

Scale for x is already present. Adding another scale for x, which will replace the existing scale.



Monthly Rent \$\$

Question 3: Part a

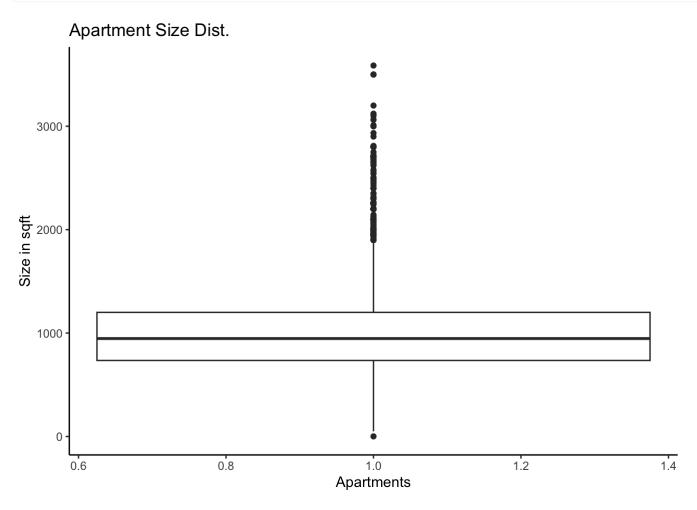
```
library(ggplot2)
library(dplyr)
craigslist <- read.csv("~/jp-garcia-131a/craigslist.csv")

craigslist %>%
  filter(size <= 60000) %>%
  ggplot(aes(x = 1, y = size)) +
```

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```
geom_boxplot() +
labs(
   title = "Apartment Size Dist. ",
   x = "Apartments",
   y = "Size in sqft"
) +
theme_classic()
```

Lab 1



Question 3: Part b

```
library(ggplot2)
library(dplyr)

filtered_listings <- craigslist %>%
    filter(location %in% c("berkeley", "oakland", "richmond", "emeryville", "albany", "el c

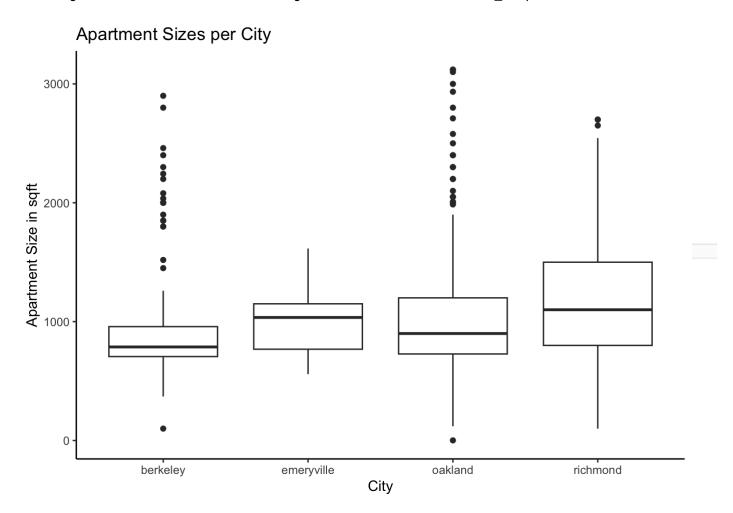
ggplot(data = filtered_listings, aes(x = location, y = size)) +
    geom_boxplot() +
    labs(
        title = "Apartment Sizes per City",
        x = "City",
        y = "Apartment Size in sqft"
```

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```
) +
theme_classic()
```

Lab 1

Warning: Removed 761 rows containing non-finite values (`stat_boxplot()`).



Intro to Probability: Question 1

```
library(ggplot2)
library(dplyr)

median_rent<- median(craigslist$price, na.rm = TRUE)

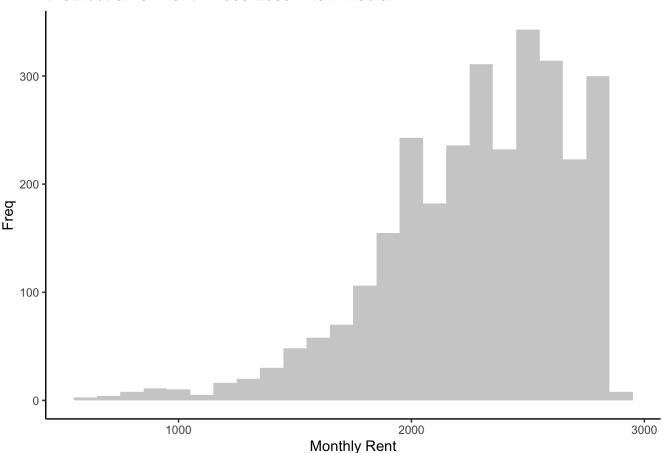
print(median_rent)</pre>
```

[1] 2865

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```
x = "Monthly Rent",
y = "Freq"
) +
theme_classic()
```

Distribution of Rent Prices Less Than Median



Question 2:

```
median_price <- median(craigslist$price)

below_median <- subset(craigslist, price < median_price)

num_under_2000 <- sum(below_median$price < 2000)

total_under_median = length(below_median$price)

prob= num_under_2000 / total_under_median

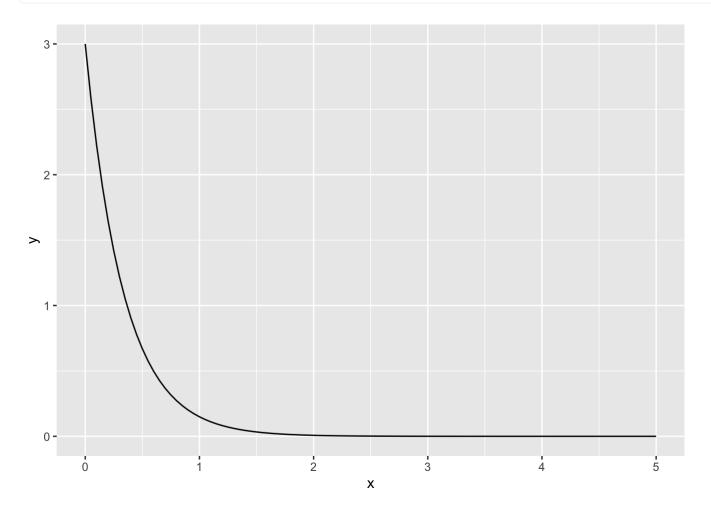
print(paste("Estimated Probability:", round(prob, 4)))</pre>
```

[1] "Estimated Probability: 0.2292"

Simulating with a Gamma Distribution: Question 2

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```
library(ggplot2)  ggplot(data.frame(x = c(0, 5)), aes(x = x)) + \\ geom_function(fun = function(x) dgamma(x, shape = 1, rate = 3))
```



Question 4:

```
Pless_than_0_1 <- pgamma(0.1, shape = 1, rate = 3)

Pgreater_1_5 <- 1 - pgamma(1.5, shape = 1, rate = 3)

total_prob <- Pless_than_0_1 + Pgreater_1_5

rounded_prob <- round(total_prob, 4)

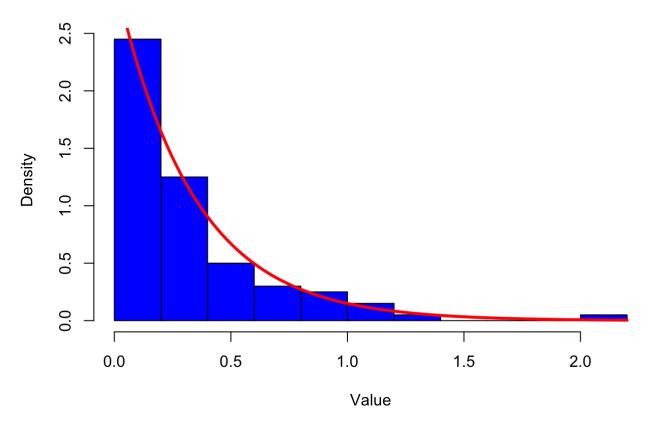
print(paste("Probability of observation being less than 0.1 or greater than 1.5:", rounded_prob.</pre>
```

[1] "Probability of observation being less than 0.1 or greater than 1.5: 0.2703"

Question 5:

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Density Curve/Histogram Overlay



Question 6:

```
count <- sum(simulation < 0.1 | simulation > 1.5)
hyp_prob <- count / length(simulation)
print(paste("Estimated probability:", round(hyp_prob, 4)))</pre>
```

[1] "Estimated probability: 0.32"

Question 7

```
print("looking at our given parameters and our sample of 100 values:
Probability of an observation < 0.1:</pre>
```

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```
With a sample of 100, this would be a reliable estimate as it would fall towards the apex Probability of an observation between 0.5 and 1.0:

Probability would be better estimated as the probability of an observation between 0.5 an Probability of an observation > 1:

As this strays away from the higher density areas of our distribution, we would need more
```

[1] "looking at our given parameters and our sample of 100 values:\n\nProbability of an observation < 0.1: \nWith a sample of 100, this would be a reliable estimate as it would fall towards the apex of our distribution \n\nProbability of an observation between 0.5 and 1.0:\nProbability would be better estimated as the probability of an observation between 0.5 and 1.0 would fall where the density is of highest in our sample \n\nProbability of an observation > 1: \nAs this strays away from the higher density areas of our distribution, we would need more data as we approach the tails of the distribution"

Distributions of Sample Data: Question 1

```
craigslist_all <- read.csv("~/jp-garcia-131a/craigslist_all.csv")

library(dplyr)

four_bed_craigslist_all <- craigslist_all %>%
    filter(brs <= 4)

craigslist <- four_bed_craigslist_all %>%
    sample_n(5876)

head(craigslist)
```

```
1 2016-09-29 09:45:00 2262 740
2 2016-09-26 12:37:00 4295 1910
3 2016-10-08 17:40:00 2229 783
4 2016-09-27 08:22:00 5500 1600
                                  3
5 2016-10-09 13:11:00 3500 1328
                                  3
6 2016-09-26 17:49:00 2250 640
                                  1
                                                              title
         Enjoy The Best Deals In San Jose 8 Weeks Free 99 Deposit
1
2
    Cambrian Beauty Sunny and Bright 4BR 3BA KB Single Family Home
3
       READY TO MOVE TODAY OneBedroom Apartment Home Turn Key Ready
4 Modern Luxury Townhome in Heart of Mill Valley Available for Rent
5
                                                    Condo for rent
6
      Upgraded Washer Dryer Pet Friendly And First Month Rent Free
                                                 location
                     link
1 /sby/apa/5804920120.html
                                           san jose south
                                   willow glen / cambrian
2 /sby/apa/5764055699.html
3 /eby/apa/5819684163.html dublin / pleasanton / livermore
4 /nby/apa/5789034890.html
                                              mill valley
```

time price size brs

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5 /eby/apa/5820624683.html

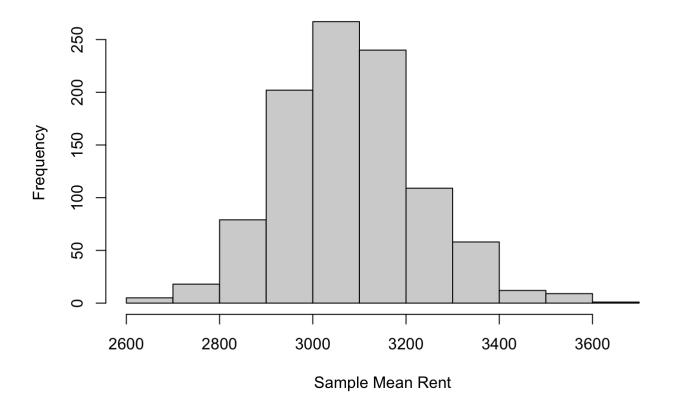
6 /sby/apa/5800823085.html san jose north

Question 2:

```
# Sample size
n <- 100
# Number of samples
samples <- 1000
sample_means <- replicate(samples, {</pre>
  sample_data <- sample_n(subset(craigslist_all, brs <= 4), n)</pre>
  mean(sample_data$price)
})
# Visualize the sampling distribution
hist(sample_means, main = "Sampling Distribution of Mean Rent", xlab = "Sample Mean Rent"
```

danville / san ramon

Sampling Distribution of Mean Rent



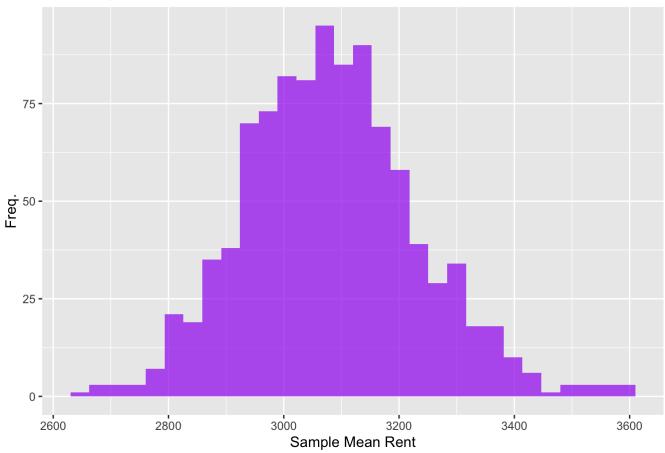
Question 3:

```
library(ggplot2)
sample_means_df <- data.frame(sample_means)</pre>
```

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```
ggplot(sample_means_df, aes(x = sample_means)) +
geom_histogram(fill = 'purple', alpha = 0.7, bins = 30) +
labs(title = "Sampling Dist. Rent", x = "Sample Mean Rent", y = "Freq.")
```

Sampling Dist. Rent



Question 4:

```
print("Extremely unsure how to go about this problem")
```

[1] "Extremely unsure how to go about this problem"

Question 5:

```
x_vector <- c(0, 1, -2) # Possible values
probs_vector <- c(1/3, 1/3, 1/3) # Associated probabilities

# Number of 'X' to sum for each observation (sample size)
m <- 10000

# Number of observations (replications)
B <- 5000

n_sums <- replicate(
B,</pre>
```

```
mean(
    sample(x = x_vector, size = m, replace = TRUE, prob = probs_vector)
)
)
```

Question 6:

```
mu <- -1/3
variance <- 14/9 / 10000
sd <- sqrt(variance)
n_sum_normal <- rnorm(5000, mean = mu, sd = sd)
summary(n_sum_normal)</pre>
```

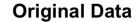
```
Min. 1st Qu. Median Mean 3rd Qu. Max. -0.3737 -0.3418 -0.3334 -0.3332 -0.3248 -0.2887
```

Question 7:

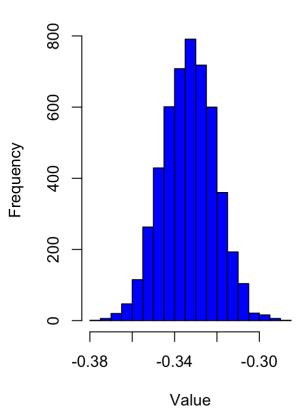
```
# Create a new window for multiple plots
par(mfrow=c(1,2))

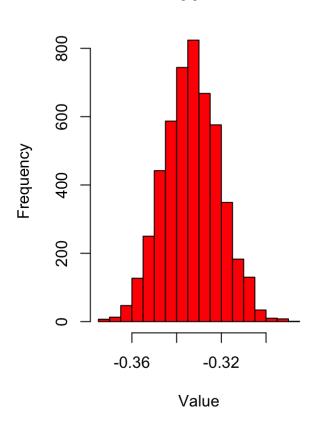
# Histogram for original data
hist(n_sums, main="Original Data", xlab="Value", ylab="Frequency", col="blue")

# Histogram for approximate data
hist(n_sum_normal, main="Normal Approximation", xlab="Value", ylab="Frequency", col="red")
```



Normal Approximation





Density Estimation: Question 1

```
library(dplyr)
heart_disease <- read.csv("~/jp-garcia-131a/heartDisease.csv")
library(dplyr)

# Assuming heartDisease is your data frame
heart_disease <- heart_disease %>%
    mutate(
        num = factor(num),
        cp = factor(cp)
      )
str(heart_disease)
```

```
'data.frame': 297 obs. of 14 variables:

$ age : num 63 67 67 37 41 56 62 57 63 53 ...

$ sex : num 1 1 1 1 0 1 0 0 1 1 ...

$ cp : Factor w/ 4 levels "1","2","3","4": 1 4 4 3 2 2 4 4 4 4 ...

$ trestbps: num 145 160 120 130 130 120 140 120 130 140 ...

$ chol : num 233 286 229 250 204 236 268 354 254 203 ...

$ fbs : num 1 0 0 0 0 0 0 0 1 ...

$ restecg : num 2 2 2 0 2 0 2 0 2 2 ...
```

```
$ thalach : num 150 108 129 187 172 178 160 163 147 155 ...
$ exang : num 0 1 1 0 0 0 0 1 0 1 ...
$ oldpeak : num 2.3 1.5 2.6 3.5 1.4 0.8 3.6 0.6 1.4 3.1 ...
$ slope : num 3 2 2 3 1 1 3 1 2 3 ...
$ ca : num 0 3 2 0 0 0 2 0 1 0 ...
$ thal : num 6 3 7 3 3 3 3 3 7 7 ...
$ num : Factor w/ 5 levels "0","1","2","3",..: 1 3 2 1 1 1 4 1 3 2 ...
```

Question 2:

```
heart_disease %>%
   select(num,cp) %>%
   table()
```

```
cp
num 1 2 3 4
0 16 40 65 39
1 5 6 9 34
2 1 1 4 29
3 0 2 4 29
4 1 0 1 11
```

```
print("This code will only select the num and cp columns from the heart_disease data fram
```

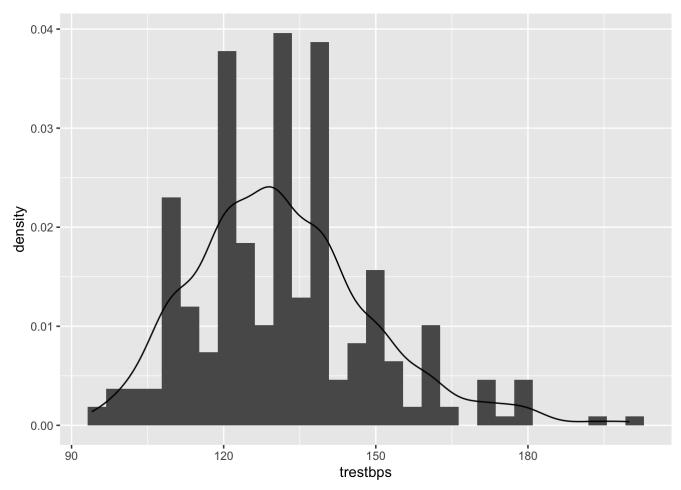
[1] "This code will only select the num and cp columns from the heart_disease data frame and the table function will give us a table summary of how many occurences there are per unique combination of num and cp"

Question 3:

```
library(ggplot2)

ggplot(heart_disease, aes(x=trestbps)) +
  geom_histogram(aes(y = ..density..), bins = 30) +
  geom_density()
```

Warning: The dot-dot notation (`..density..`) was deprecated in ggplot2 3.4.0. i Please use `after_stat(density)` instead.



print("Looking at the shape of the distribution we are able to see that there is a right

[1] "Looking at the shape of the distribution we are able to see that there is a right skewed but largely centered around values 120–125. The blood pressure seems to be mostly normal while there is a proportion that seems to hover in the high pressure" Question 4:

Age Density Estimate by Diagnosis

