

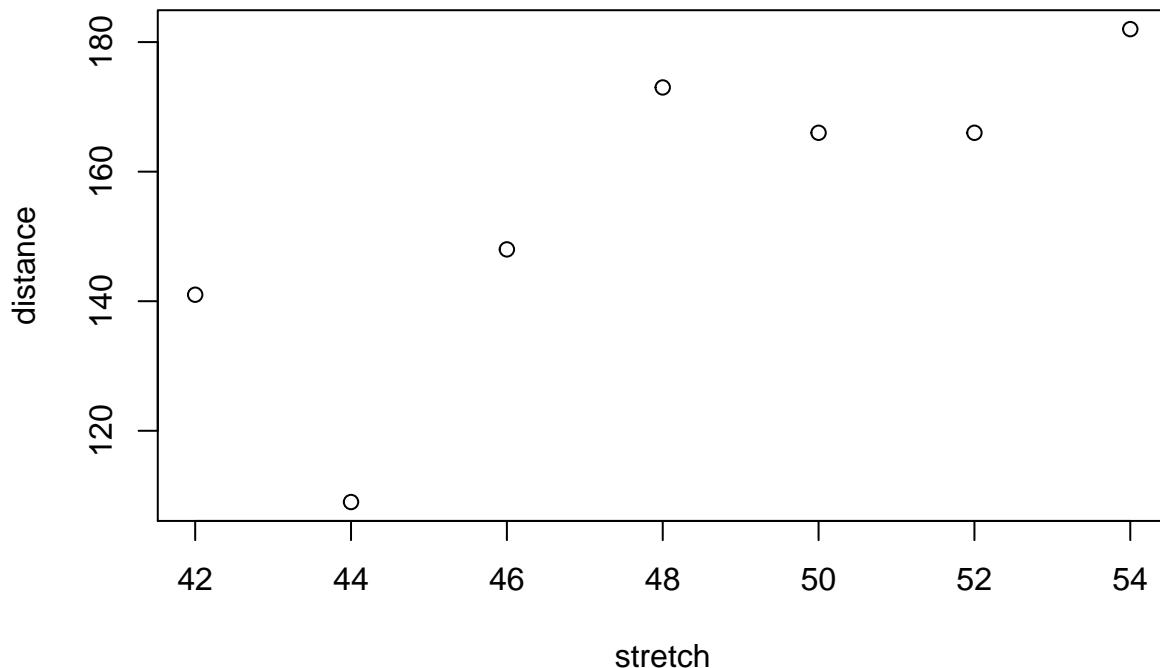
# Homework 00 Solutions

## Chapter 1 Exercises

```
load("../usingR.RData")
```

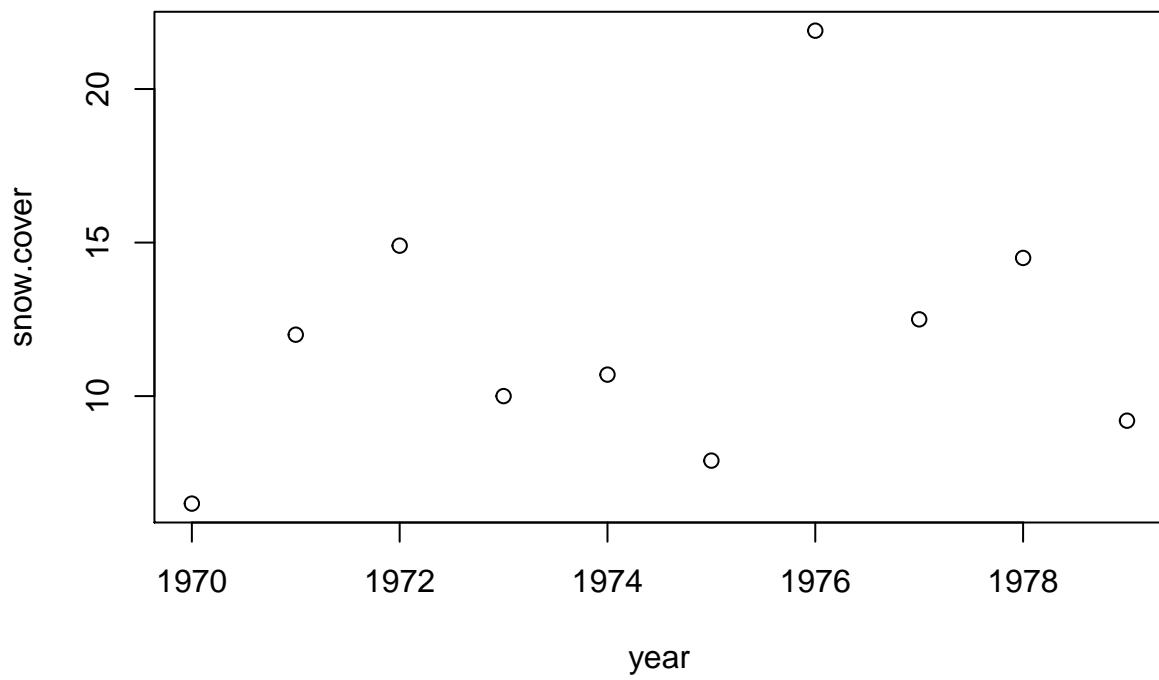
### Exercise 1

```
plot(distance ~ stretch, data = elasticband)
```



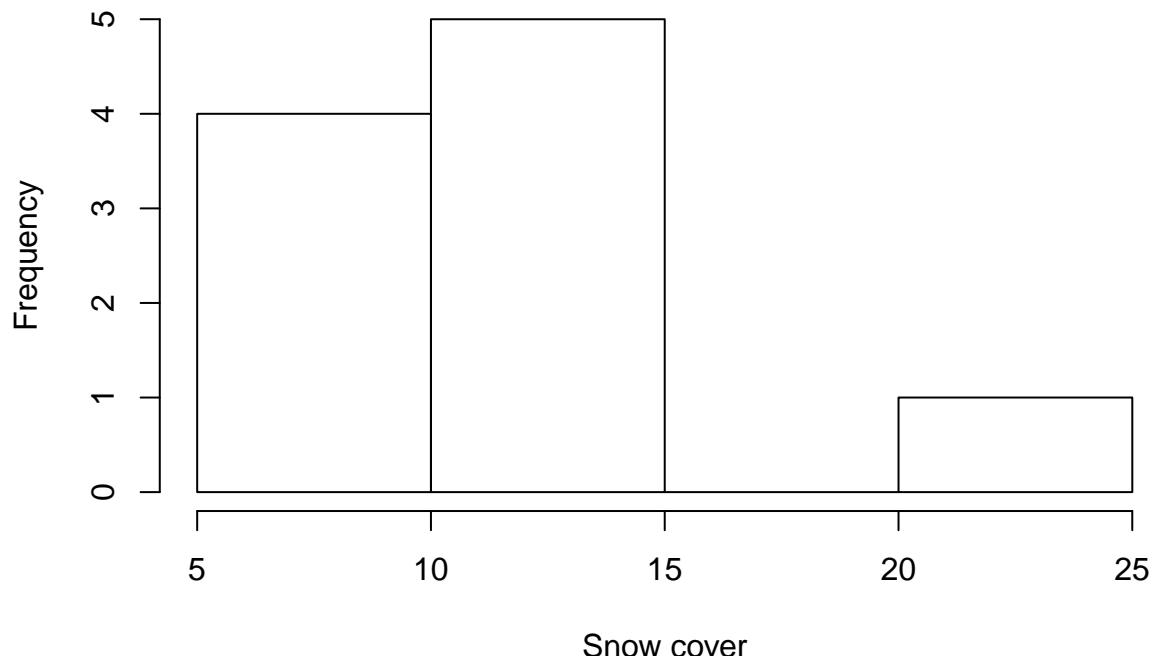
### Exercise 2

```
# i.  
snow <- data.frame(year = 1970:1979,  
                     snow.cover = c(6.5, 12, 14.9, 10, 10.7, 7.9, 21.9, 12.5, 14.5, 9.2))  
  
# ii.  
plot(snow.cover ~ year, data = snow)
```

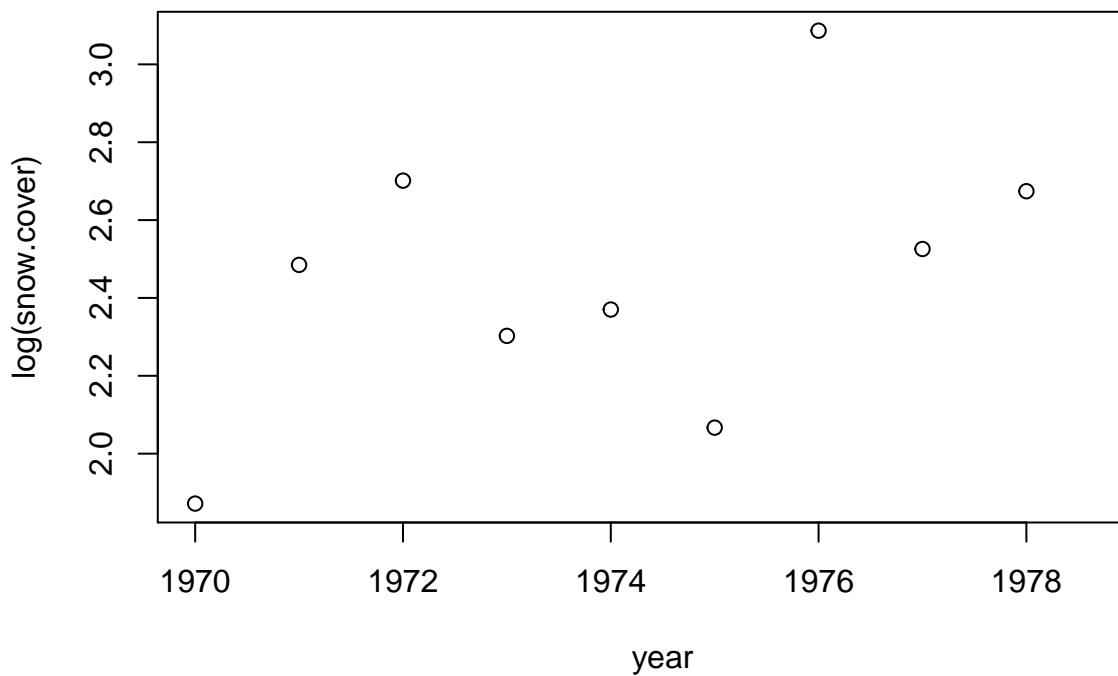


```
# iii.  
hist(snow$snow.cover, xlab = "Snow cover", main = "Histogram of snow cover")
```

**Histogram of snow cover**

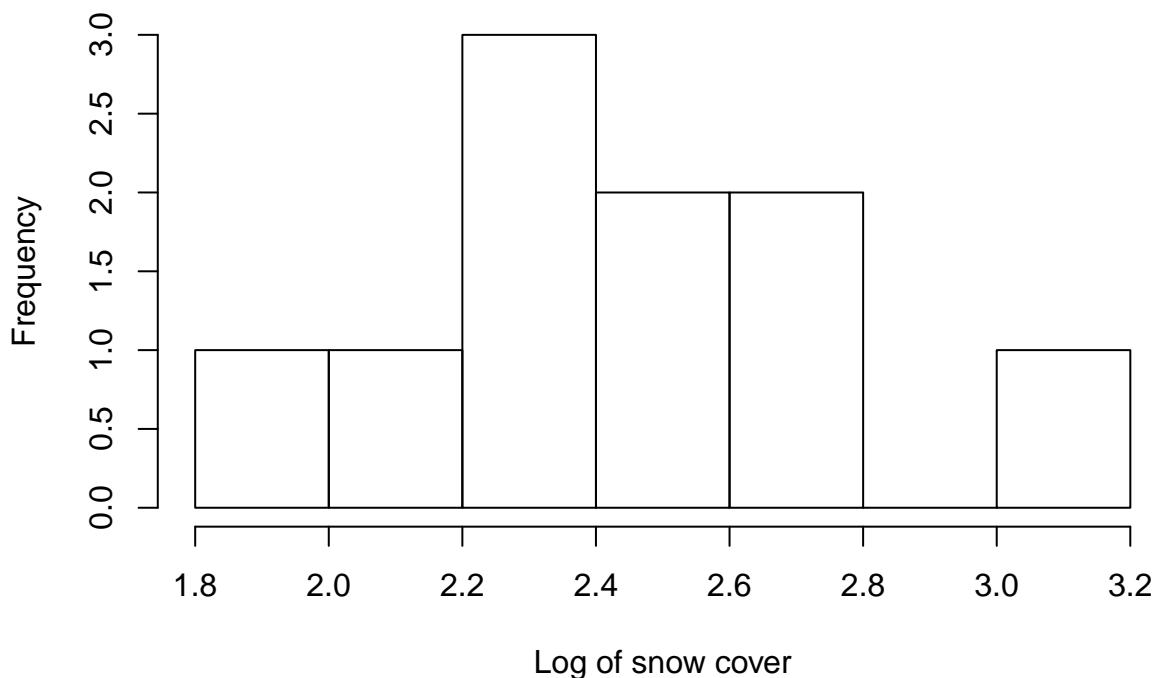


```
# iv.  
plot(log(snow.cover) ~ year, data = snow)
```



```
hist(log(snow$snow.cover), xlab = "Log of snow cover", main = "Histogram of snow cover")
```

**Histogram of snow cover**



### Exercise 3

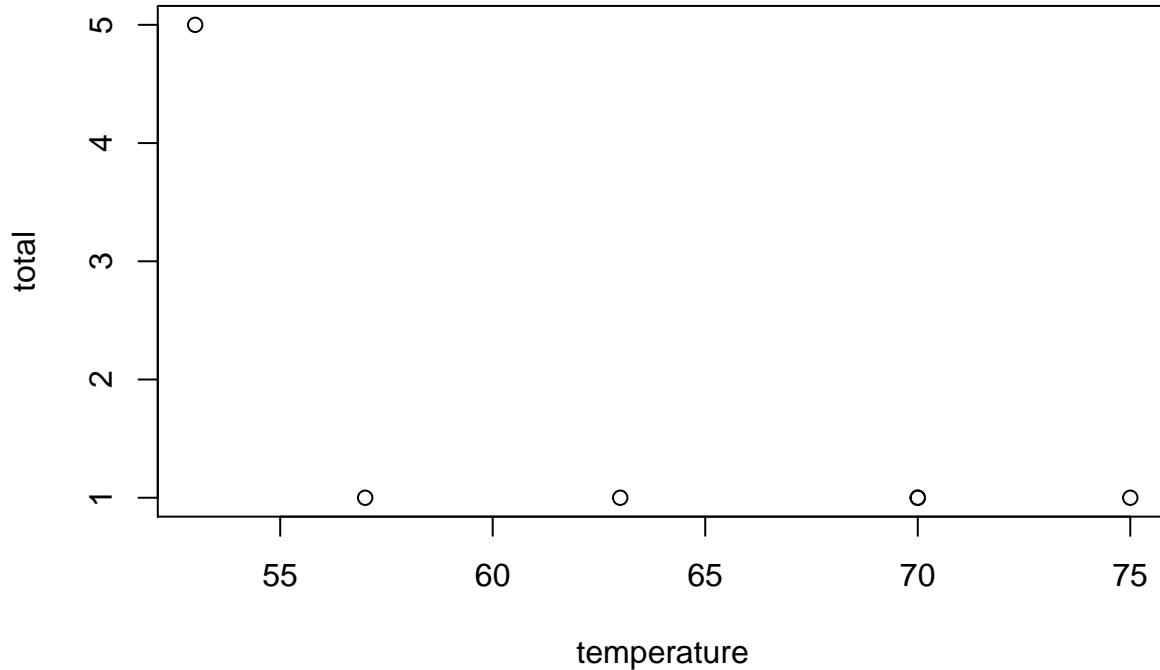
```
orings <- data.frame(temperature = c(53, 57, 63, 70, 70, 75),  
                     erosion = c(3, 1, 1, 1, 1, 0),
```

```

blowby = c(2, 0, 0, 0, 0, 2),
total = c(5, 1, 1, 1, 1, 1))

plot(total ~ temperature, data = orings)

```



## Chapter 2 Exercises

### Exercise 1

```

# 2.1(a)
answer <- 0
for (j in 3:5) {answer <- j + answer}
print(answer)

```

```

## [1] 12
# 2.1(b)
answer <- 10
for (j in 3:5) {answer <- j + answer}
print(answer)

```

```

## [1] 22
# 2.1(c)
answer <- 10
for (j in 3:5) {answer <- j*answer}
print(answer)

```

```

## [1] 600

```

## Exercise 2

```
?prod  
prod(c(10, 3, 4, 5))  
  
## [1] 600
```

## Exercise 5

```
radii <- 3:20  
spheres <- data.frame(radius = radii, volume = 4*pi*radii^3/3)  
spheres  
  
##      radius      volume  
## 1        3    113.0973  
## 2        4    268.0826  
## 3        5    523.5988  
## 4        6    904.7787  
## 5        7   1436.7550  
## 6        8   2144.6606  
## 7        9   3053.6281  
## 8       10   4188.7902  
## 9       11   5575.2798  
## 10      12   7238.2295  
## 11      13   9202.7721  
## 12      14  11494.0403  
## 13      15  14137.1669  
## 14      16  17157.2847  
## 15      17  20579.5263  
## 16      18  24429.0245  
## 17      19  28730.9120  
## 18      20  33510.3216
```

When you type `pi` in R and hit `Enter`, R prints the value of  $\pi$  to some number of digits.

## Exercise 6

```
# Identify which columns are factors  
factor_columns <- sapply(tinting, is.factor)  
factor_columns  
  
##    case     id     age     sex     tint target     it     csoa    agegp  
##    FALSE    FALSE    FALSE    TRUE    TRUE    TRUE  FALSE  FALSE    TRUE  
  
# Get levels of factor columns  
sapply(tinting[factor_columns], levels)  
  
## $sex  
## [1] "f" "m"  
##  
## $tint  
## [1] "no" "lo" "hi"  
##
```

```

## $target
## [1] "locon" "hicon"
##
## $agegp
## [1] "Younger" "Older"
# Check which columns are ordered factors
sapply(tinting, is.ordered)

##   case     id    age    sex    tint target      it    csoa    agegp
##   FALSE  FALSE  FALSE  FALSE    TRUE  FALSE  FALSE  FALSE  FALSE

```

## Chapter 3 Exercises

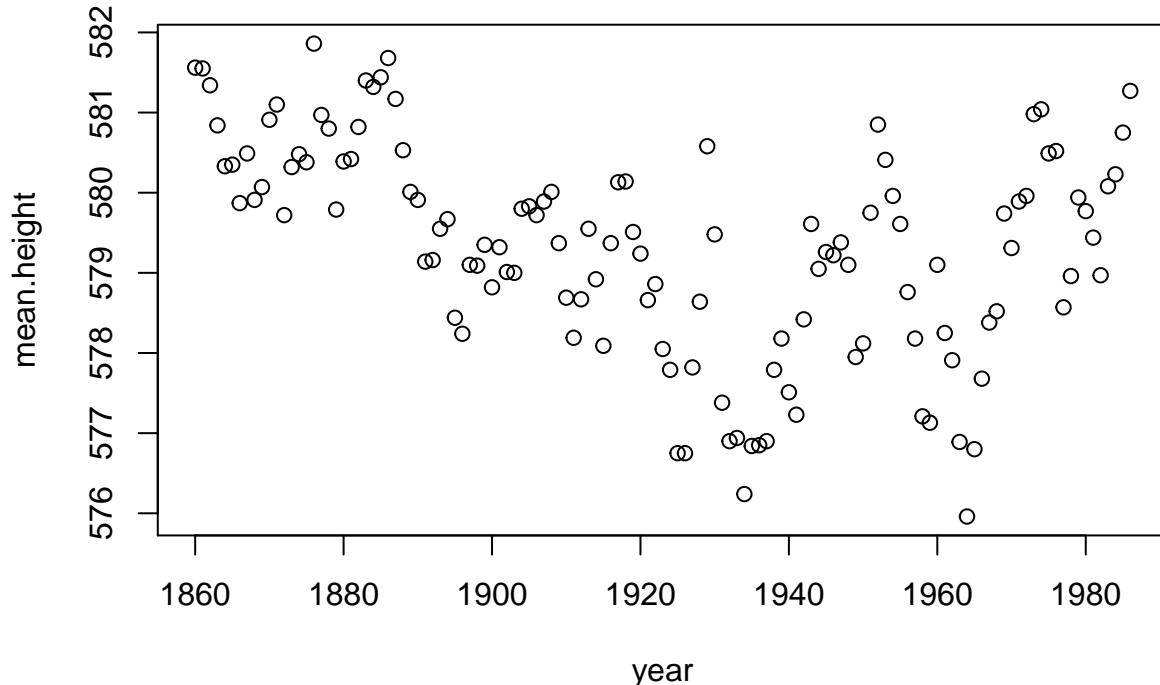
### Exercise 1

```

# 3.1(a)
plot(mean.height ~ year, data = huron)

# 3.1(b)
# Note: These labels won't necessarily show up in your saved output, which is fine.
identify(huron$year, huron$mean.height, labels = huron$year)

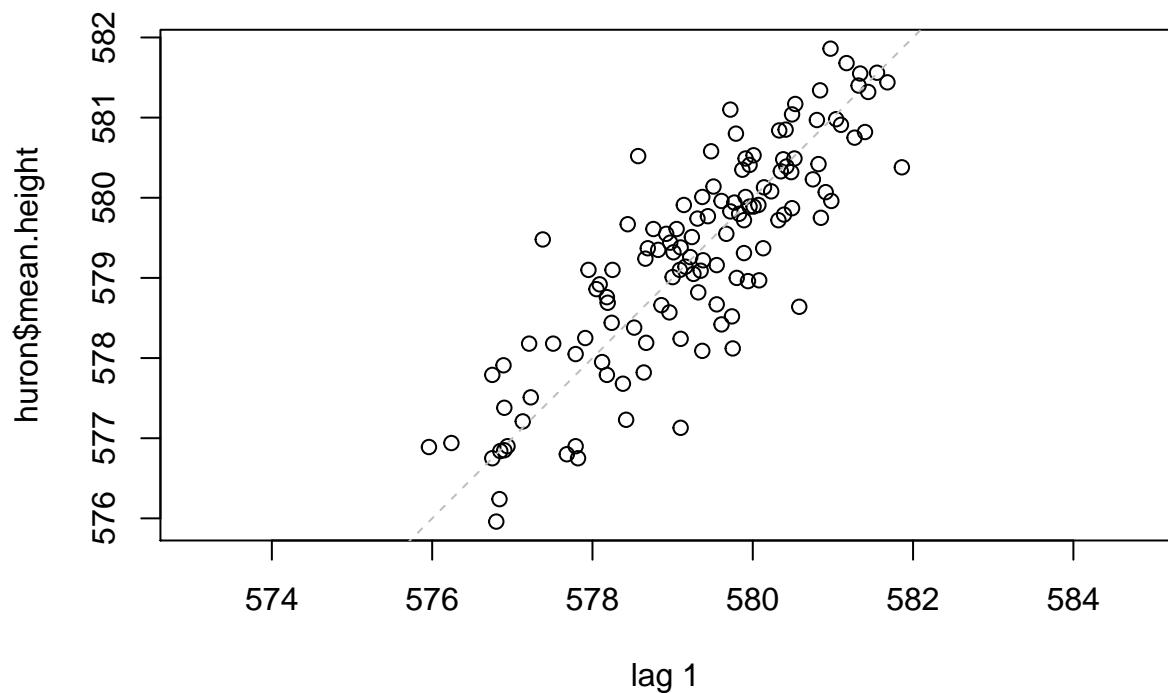
```



```

## integer(0)
# 3.1(c)
# Note: If you read the docs for lag.plot(), with ?lag.plot or help(lag.plot),
# you'll see that you can't change certain graphical parameters, including xlab and ylab.
lag.plot(huron$mean.height, do.lines = FALSE)

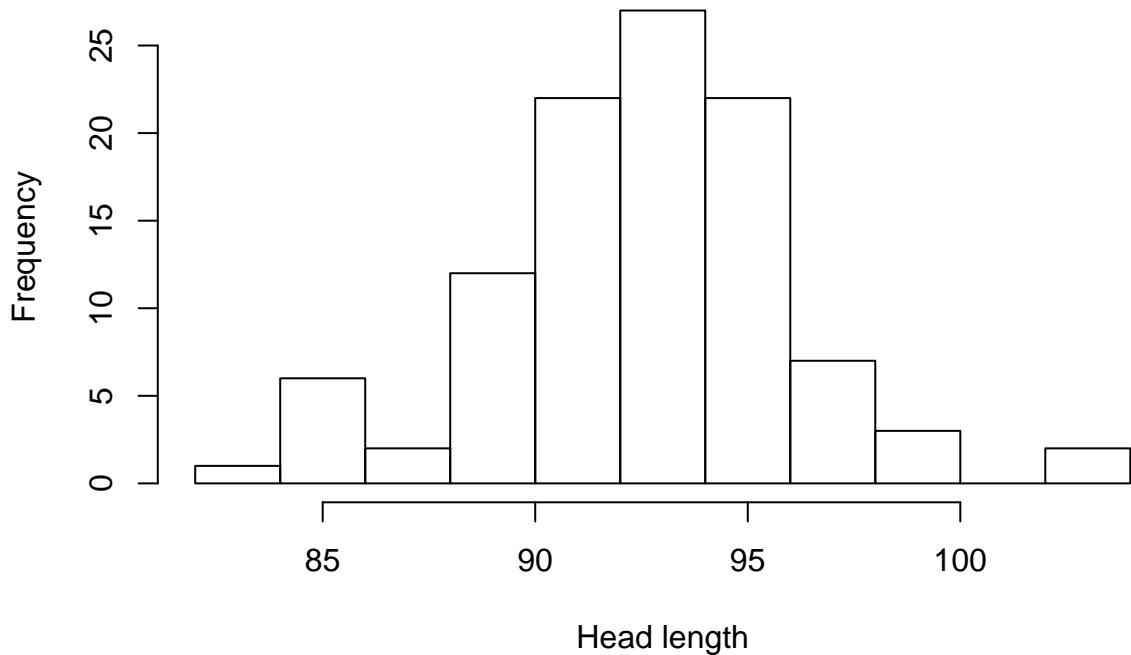
```



### Exercise 3

```
# 3.3(a)  
hist(possum$hdlngth, xlab = "Head length")
```

**Histogram of possum\$hdIngth**

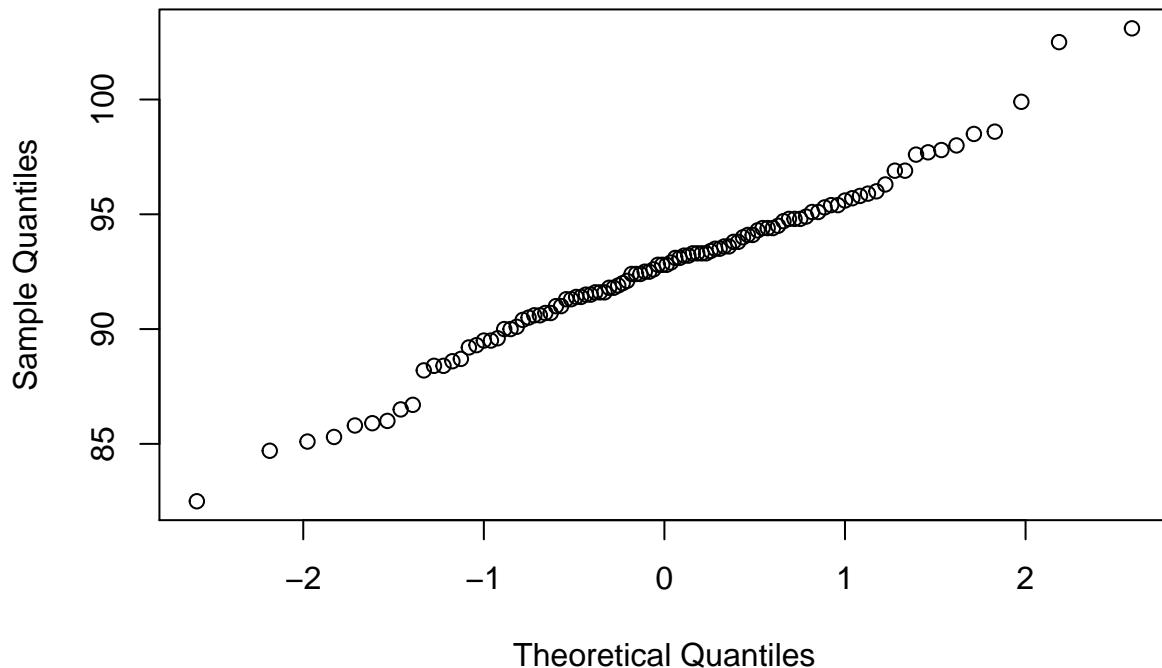


```
# 3.3(b)
# Note: the code in the problem is incorrectly given as stem(qqnorm(possum$hdlength))
stem(possum$hdlength)

##
## The decimal point is at the |
##
## 82 | 5
## 83 |
## 84 | 7
## 85 | 1389
## 86 | 057
## 87 |
## 88 | 24467
## 89 | 23556
## 90 | 001456677
## 91 | 00334455666889
## 92 | 014445568889
## 93 | 112233334556688
## 94 | 0113444578889
## 95 | 113446789
## 96 | 0399
## 97 | 678
## 98 | 056
## 99 | 9
## 100 |
## 101 |
## 102 | 5
## 103 | 1

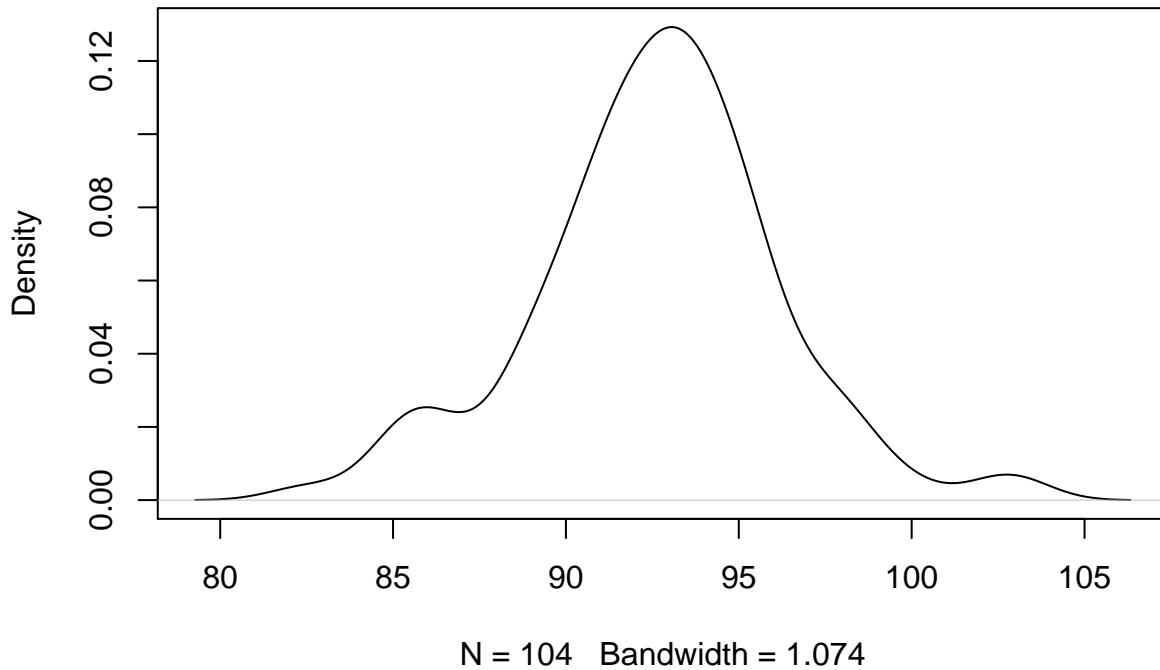
# 3.3(c)
qqnorm(possum$hdlength)
```

## Normal Q-Q Plot



```
# 3.3(d)  
plot(density(possum$hdlnth), main = "Estimated density of possum head length")
```

## Estimated density of possum head length

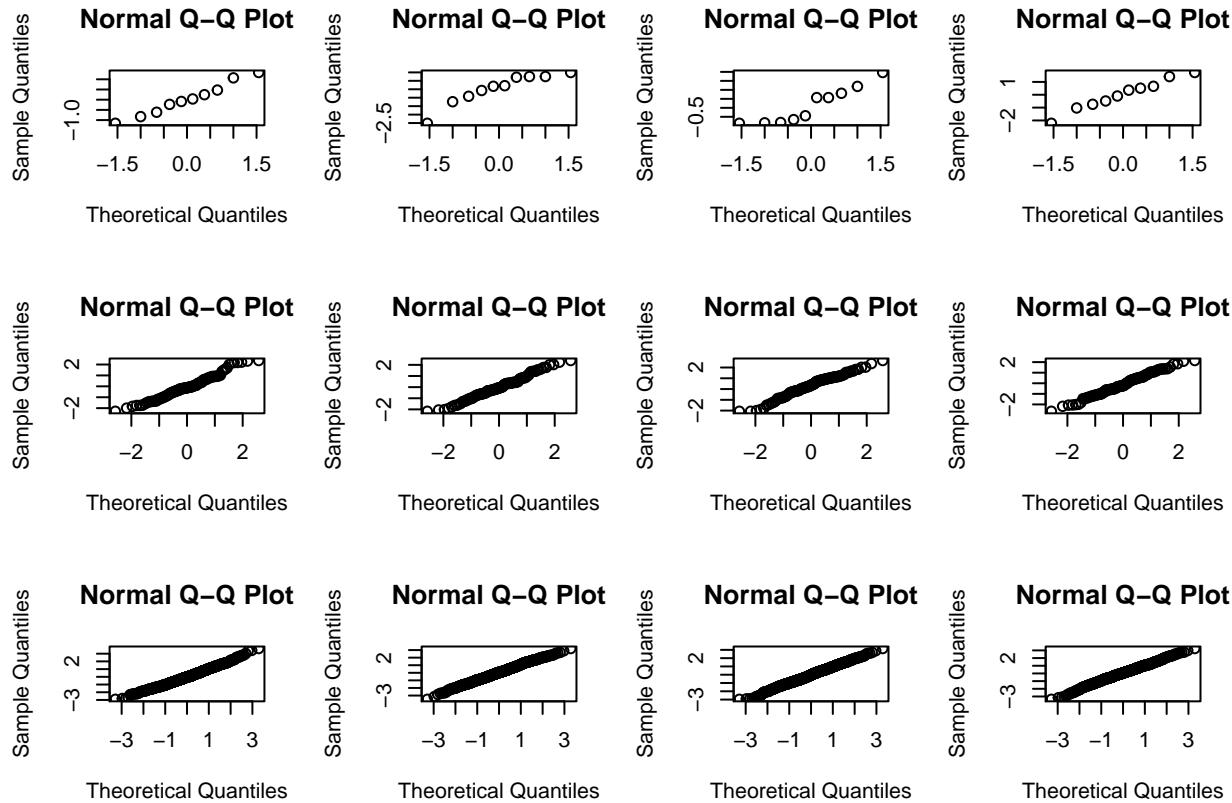


These displays provide slightly different types of information. The stem and leaf plot is kind of a sideways histogram, with the bin widths fixed at 1, but it allows you to see all the individual data points. The

histogram provides more flexibility in that bin widths are adjustable, but it is sensitive to the bin width chosen. The density plot is a smoothed version of the histogram, which may give a better sense of the shape of the distribution, but it is sensitive to the parameters of kernel used to estimate the curve. The normal QQ plot makes it easy to see deviations from normality, but it is hard to see what the distribution is if it is not normal. The other displays make it much easier to see the shape of the distribution.

## Exercise 5

```
par(mfrow = c(3, 4))
for (i in 1:4) {
  qqnorm(rnorm(10))
}
for (i in 1:4) {
  qqnorm(rnorm(100))
}
for (i in 1:4) {
  qqnorm(rnorm(1000))
}
```



```
# Typically you'll want to reset the grid after changing it.
par(mfrow = c(1, 1))
```

The larger the sample size, the more representative the sample (i.e. the closer to normal), and hence the straighter the line in the QQ plot.