1. a

Source:

> x2 = rnorm(9, 60, 7)

> g = replicate(100, Cl.t.test(x2))

> g[1:10]

[1] 56.07445 66.98801 56.07445 66.98801 56.07445 66.98801 56.07445 66.98801

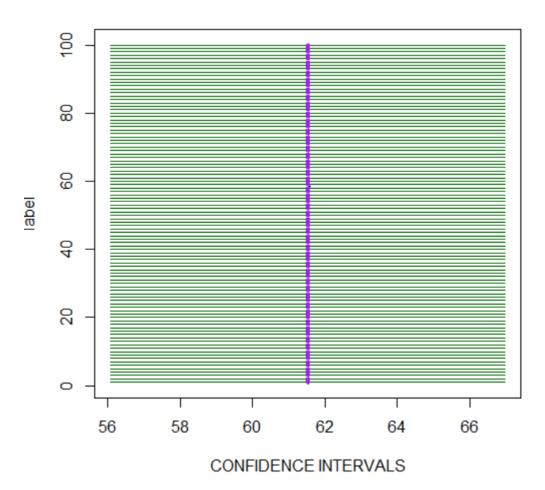
[9] 56.07445 66.98801

> mu = mean(x2)

b.

> plotCl(g,mu=mu)

100 % of these confidence intervals contain 61.53123



c. 100 percent of the intervals contain the population mean

2. a

Source: > y = read.table2("EXER4_41.txt")

```
> y
Output: V1 V2 V3
1 1278 Y
2 2 192 Y
3 3310 Y
4 4 94 N
5 5 86 Y
6 6335 Y
7 7310 N
8 8 290 Y
9 9 221 Y
10 10 168 Y
11 11 188 N
12 12 212 N
13 13 92 Y
14 14 56 Y
15 15 142 Y
16 16 37 Y
17 17 186 N
18 18 221 Y
19 19 219 N
20 20 305 Y
b.
Source:
> f = c(y[,2])
> N = 362
> mean(f) * N
Output: [1] 71350.2
c. > fpc0 = (1-n/N)
> b = 2* N* sd(f)* sqrt(1/n* fpc0)
> b
Output: [1] 14296.9
> mydesign = svydesign(~1, fpc = rep(N,n), data = g)
> h = svytotal(~f,mydesign)
total SE
f 71350 7148.5
> b2 = SE(h) * 2
> b2
Output: f 14296.9
<mark>e.</mark>
```

Source:

```
> d = (4800^2)/(4*362^2)
> s2 = sum((f-mu)^2)*(1/19)
> N*s2/(N-1)*d + s2
Output: [1] 372108.1
3a.
(-6, -4)
(-6,3)
(-6,9)
(-4,3)
(-4,9)
(3,9)
b.
> var(c(-6,-4))
[1] 2
> var(c(-6,3))
[1] 40.5
> var(c(-6,9))
[1] 112.5
> var(c(-4,3))
[1] 24.5
> var(c(-4,9))
[1] 84.5
> var(c(3,9))
[1] 18
c.
>(2*1/6) + (40.5*1/6) + (112.5 * 1/6) + (24.5*1/6) + (84.5*1/6)+(18*1/6)
d. > ((y1^2*1/6) + (y2^2*1/6) + (y3^2*1/6) + (y4^2*1/6) + (y5^2*1/6) + (y6^2*1/6))
[1] 12
> mu = ((y1*1/6) + (y2*1/6) + (y3*1/6) + (y4*1/6) + (y5*1/6) + (y6*1/6))
> 12 - mu^2
[1] 11.75
```

e. The value in part c should be greater than the true population variance because our population size is so low, this creates for a greater fpc and ultimately results in a very high variance between the values. If, N was huge compared to n, we wouldn't have to worry about the variance as much, as it would most likely be close to the population variance.

4.

D

5.

F