

This is an example of Parametric Bootstrapping to find the approximate standard error and bias of a statistic. We use the Bar Soap data encountered before. A QQ-Plot shows that the sample is reasonably normal, so we assume it is taken from the distribution $N(\mu, \sigma)$. The sample mean \bar{x} and standard deviation s are estimators for μ and σ . We are interested in the standard error of the sample standard deviation. We know that the sample variance s^2 is an unbiased estimator for σ^2 , but because we make a nonlinear transformation, s is not necessarily an unbiased estimator for σ . In principle, we could derive the theoretical standard error knowing that it is the square root of the sample variance which satisfies a chi-squared distribution.

But, now with availability of computing power, it is easier just to simulate and measure! The idea is to take a random sample of size n from $N(\bar{x}, s)$, call it x_1^*, \dots, x_n^* , take its standard deviation and call that the first bootstrapped estimate s_1^* . We repeat B times to get simulated sd's $s_1^*, s_2^*, \dots, s_B^*$ and then the mean and standard deviations give the bootstrapped estimates

$$\bar{s}^* = \frac{1}{B} \sum_{i=1}^B s_i^*; \quad S_{s^*} = \sqrt{\frac{1}{B-1} \sum_{i=1}^B (s_i^* - \bar{s}^*)^2}.$$

S_{s^*} is an estimate for the standard error σ_s . However, we have to correct for bias. Denote bias of the sample standard deviation by

$$b_s = E(s) - \sigma.$$

An estimate for the bias is given by

$$\hat{b}_s = \bar{s}^* - s$$

which is the difference of the performance of the sample standard deviations against the bootstrap population standard deviation s . If \hat{b}_s is negative then the sample standard deviation is underestimating the population sd. Since we are simulating a population whose parameters are close to the background $N(\mu, \sigma)$, the bias in the bootstrap sample will be close to the background sample. Thus the bias corrected estimate of the background sd is

$$s^{**} \approx s - \hat{b}_s.$$

The bias also influences the standard error. Indeed, the square of the standard error is

$$\begin{aligned} \sigma_s^2 &= E([S - \sigma]^2) \\ &= E([S - E(S) + E(S) - \sigma]^2) \\ &= E([S - E(S)]^2 + 2[S - E(S)][E(S) - \sigma] + [E(S) - \sigma]^2) \\ &= E([S - E(S)]^2) + 2E(S - E(S))[E(S) - \sigma] + E([E(S) - \sigma]^2) \\ &= E([S - E(S)]^2) + 0 + b_s^2 \\ &\approx S_{s^*}^2 + (\bar{s}^* - s)^2. \end{aligned}$$

Thus the bias corrected approximate standard error is

$$se^{**} = \sqrt{S_{s^*}^2 + \hat{b}_s^2}.$$

R Session:

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[R.app GUI 1.34 (5589) i386-apple-darwin9.8.0]

[Workspace restored from /home/1004/ma/treibergs/.RData]

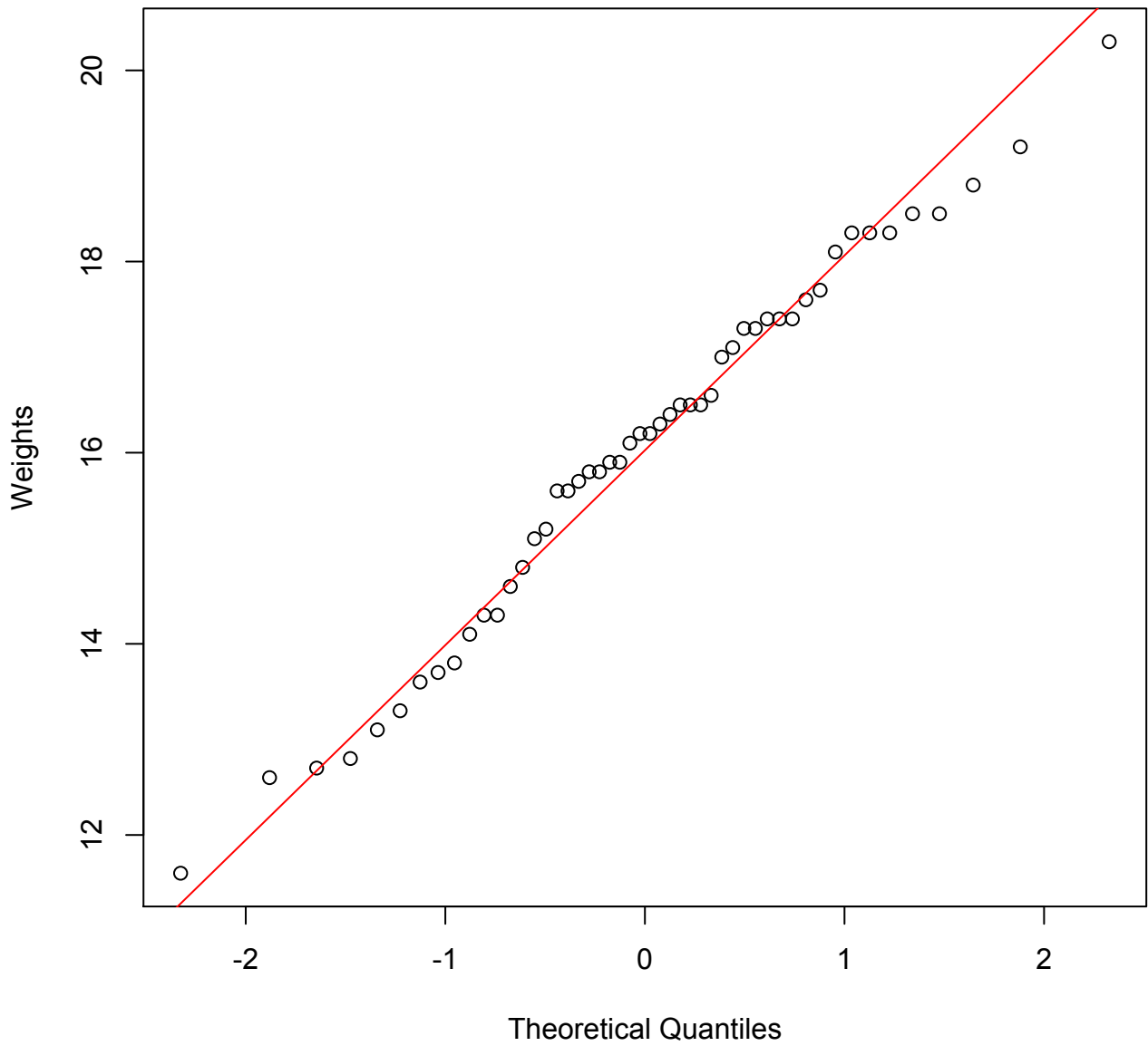
```
> tt <- read.table("M3073BarSoapData.txt",header=TRUE)
> attach (tt)
> tt
```

```
  Weights
1      11.6
2      12.6
3      12.7
4      12.8
5      13.1
6      13.3
7      13.6
8      13.7
9      13.8
10     14.1
11     14.3
12     14.3
13     14.6
14     14.8
15     15.1
16     15.2
17     15.6
18     15.6
19     15.7
20     15.8
21     15.8
22     15.9
```

23	15.9
24	16.1
25	16.2
26	16.2
27	16.3
28	16.4
29	16.5
30	16.5
31	16.5
32	16.6
33	17.0
34	17.1
35	17.3
36	17.3
37	17.4
38	17.4
39	17.4
40	17.6
41	17.7
42	18.1
43	18.3
44	18.3
45	18.3
46	18.5
47	18.5
48	18.8
49	19.2
50	20.3

```
> ##### CHECK NORMALITY OF THE BACKGROUND POPULATION #####  
> qqnorm(Weights, ylab="Weights")  
> qqline(Weights, col=2)  
> # M3074SoapBoot1.pdf  
> # Looks Pretty normal
```

Normal Q-Q Plot



```

> ##### FIND DISTRIBUTION PARAMETERS #####
> xbar <- mean(Weights); xbar
[1] 16.034
> s <- sd(Weights); s
[1] 1.948585
>
> ##### BOOTSTRAP SIMULATION FROM APPROXIMATE BACKGROUND POP. #####
>
> # Sample size n = 50, No. Bootstrap Samples B = 10,000
>
> sstar <- replicate(10000,sd(rnorm(50,xbar,s)))
>
> # The bootstrap sample mean gives biased samples.
> sstarbar <- mean(sstar); sstarbar
[1] 1.937850
>
> # The bootstrap sd gives the biased approximate se
> sestar <- sd(sstar); sestar
[1] 0.1953213
>
> # The approximate bias from the bootstrap sample.
> bias <- sstarbar-s; bias
[1] -0.01073481
>
> # Correct the estimated sd of the background population
> s-bias
[1] 1.95932
>
> # Correct the estimated se of the bootstrapped sample
>
> sestarstar <- sqrt(sestar^2 + bias^2); sestarstar
[1] 0.1956161
>
>
> ss <- seq(.7,3.8,1/57)
> sigmahathat <- s-bias
> lines(ss, dnorm(ss,xbar,s),col=4)
> lines(ss, dnorm(ss,s,sestar),col=4)
> lines(ss, dnorm(ss,sigmahathat,sestarstar),col=3)
>
> ##### HISTOGRAM OF BOOTSTRAPPED SAMPLE s* WITH NORMAL SUPERIMPOSED #####
>
> hist(sstar, xlab= "s*", main = paste("Histogram of Bootstrapped sd with xbar =",
+ xb, ", s =", sb), col = "gold", freq = F)
> # Plot normals with bootstrapped mean and sd
> # Uncorrected N(s,sestar) Corrected N(sigmahathat,sestarstar)
> lines(ss, dnorm(ss,s,sestar),col=4)
> lines(ss, dnorm(ss,sigmahathat,sestarstar),col=2)
> legend(1.1, 1.9, legend = c("N(s,se*)\nNo Correction\n",
+ "N(s**,se**)\nBias Corrected\n"), fill = c(4,2), bg = "white")
> # M3074SoapBoot2.pdf

```

Histogram of Bootstrapped sd with $\bar{x} = 16.034$, $s = 1.949$

