

1) $\mu_0 = 15$ mg/day

Conjectured true population mean, $\mu_A = 17$ mg/day

Hypothesis:

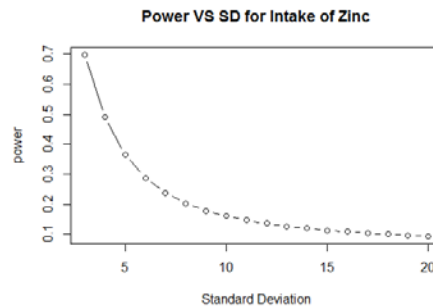
$$H_0 : \mu \leq \mu_0 : \mu \leq 15$$

$$H_A : \mu > \mu_0 : \mu > 15$$

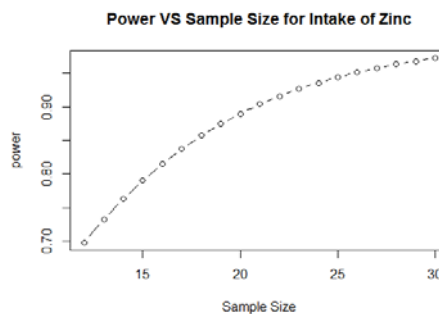
$n = 12$, $\alpha = 0.05$, conjectured s.d $\sigma = 3$ mg/day

A) power = 0.6981908 [See appendix for R code and Calculation]

B) If the sample deviation was larger, the power would be lower than that of the power that was calculated in part A. Following graph shows the Power ~ SD relation for the data.



C) If the sample size was larger, power would increase. The following graph shows the Power ~ Sample size relation for Zinc data.

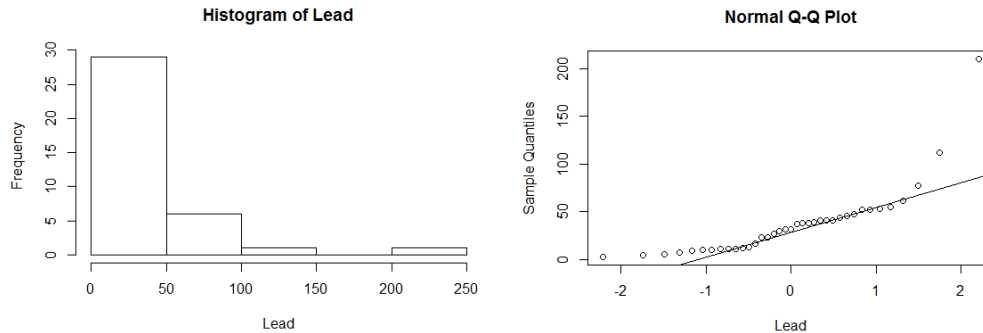


D) If $\alpha = 0.10$, power = 0.8260609 [See Appendix for calculation], which is higher than the power calculated in part A.

E) For $\mu_A = 16$ mg/day, power = 0.2874441 [See Appendix for calculation], which is lower than the power calculated in part A.

F) For Power = 0.9, sample size = 21 (rounded up) [See Appendix for calculation].

2) A)



Shapiro-Wilk normality test p-value = $1.928e-07 < 0.05$. [See appendix]
So, the null hypothesis (lead sample data is normally distributed) can be rejected.

The result of the shapiro-wilk test matches the histogram and normal Q-Q plot. From the Histogram we can see that the lead sample data is skewed right, and from the normal Q-Q plot we see that the plot is not a straight line.

B) Mean = 37.24324

Median = 32

C) $H_0 : M = 30$

$H_A : M \neq 30$

From the sign test, we obtain,
 s (#of values > 30) = 22

p-value = 0.6177

As P value $> \alpha = 0.05$, we can not reject H_0 .

We do not have enough evidence to reject the population median = 30 with 95% confidence.

D)

95% confidence interval (using Upper Archieved CI) is (17.0000, 41)

E)

$H_0 : \mu = \mu_0 : \mu = 30$

$H_A : \mu \neq \mu_0 : \mu \neq 30$

p-value = 0.2431 $> \alpha = 0.05$. So we can not reject H_0 .

We do not have enough evidence to reject that the population mean = 30 with 95% confidence.

F)

95 percent confidence interval: (24.86550, 49.62099)

G) Studentized confidence interval mean= (27.46, 57.19)

H) Assuming the cumulative lead exposure is of interest, the mean would be of more interest.

APPENDIX

#QUESTION 1

```
> #1A
> power.t.test(n=12, delta=2, sd=3, sig.level = 0.05, type = "one.sample", alternative = "one.sided")
```

One-sample t test power calculation

```
      n = 12
    delta = 2
      sd = 3
sig.level = 0.05
  power = 0.6981908
alternative = one.sided
```

```
>
> #1B
> testSD<- seq(3, 20, 1)
> powerVal1B_DiffSD <- power.t.test(n=12, delta=2, sd=testSD, sig.level = 0.05, type = "one.sample", alternative = "one.sided")
> plot(powerVal1B_DiffSD$power ~ testSD, type = "b", xlab = "Standard Deviation", ylab = "power", main = "Power VS SD for Intake of Zinc")
>
> #1C
> testSampleSize <- seq(12, 30, 1)
> powerVal1C_Diffn <- power.t.test(n=testSampleSize, delta=2, sd=3, sig.level = 0.05, type = "one.sample", alternative = "one.sided")
> plot(powerVal1C_Diffn$power ~ testSampleSize, type = "b", xlab = "Sample Size", ylab = "power", main = "Power VS Sample Size for Intake of Zinc")
>
> #1D
> power.t.test(n=12, delta=2, sd=3, sig.level = 0.10, type = "one.sample", alternative = "one.sided")
```

One-sample t test power calculation

```
      n = 12
    delta = 2
      sd = 3
sig.level = 0.1
  power = 0.8260609
alternative = one.sided
```

```
>
> #1E
> power.t.test(n=12, delta=1, sd=3, sig.level = 0.05, type = "one.sample", alternative = "one.sided")
```

One-sample t test power calculation

```
      n = 12
    delta = 1
```

```
sd = 3
sig.level = 0.05
power = 0.2874441
alternative = "one.sided"

>
> #1F
> power.t.test(delta=2, sd=3, p=0.9, sig.level = 0.05, type = "one.sample", alternative = "one.sided")
```

One-sample t test power calculation

```
n = 20.69914
delta = 2
sd = 3
sig.level = 0.05
power = 0.9
alternative = "one.sided"
```

#QUESTION2

```
> hist(DataHW4_2$X.Lead, xlab = "Lead", main = "Histogram of Lead")
> qqnorm(DataHW4_2$X.Lead, xlab = "Lead")
> qqline(DataHW4_2$X.Lead, )
> shapiro.test(DataHW4_2$X.Lead.)
```

Shapiro-Wilk normality test

```
data: DataHW4_2$X.Lead.
W = 0.69693, p-value = 1.928e-07
```

```
> #B
> mean(DataHW4_2$X.Lead.)
[1] 37.24324
> median(DataHW4_2$X.Lead.)
[1] 32
>
> #C
> hist(DataHW4_2$X.Lead, xlab = "Lead", main = "Histogram of Lead")
> summary(DataHW4_2$X.Lead.)
  Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
  3.00  11.00   32.00   37.24  46.00  210.00
> sort(DataHW4_2$X.Lead.)
[1] 3 5 6 7 9 10 10 11 11 11 12 13 17 23 23 27 30 32 32
[20] 37 38 38 39 41 41 41 44 46 48 52 52 53 55 62 77 112 210
> library(BSDA)
> SIGN.test(DataHW4_2$X.Lead, md=30)
```

One-sample Sign-Test

```
data: DataHW4_2$X.Lead.
s = 20, p-value = 0.6177
alternative hypothesis: true median is not equal to 30
95 percent confidence interval:
 17.34363 41.00000
sample estimates:
median of x
```

32

	Conf. Level	L. E. pt	U. E. pt
Lower Achi eved CI	0. 9011	23. 0000	41
Interpol ated CI	0. 9500	17. 3436	41
Upper Achi eved CI	0. 9530	17. 0000	41

```
> t.test(DataHW4_2$X. Lead. , mu=30)
```

One Sample t-test

```
data: DataHW4_2$X. Lead.
t = 1.1868, df = 36, p-value = 0.2431
alternative hypothesis: true mean is not equal to 30
95 percent confidence interval:
 24.86550 49.62099
sample estimates:
mean of x
 37.24324
```

2G

```
> mean.fun <- function(d,i)
+ {
+   m <- mean(d[i])
+   n <- length(i)
+   v <- (n-1)*var(d[i])/n^2
+   c(m,v)
+ }
> set.seed(7255)
> resultsHW4_2F <- boot(data=DataHW4_2$X. Lead. , mean.fun, R=1000)
> boot.ci(resultsHW4_2F, type="all")
```

BOOTSTRAP CONFIDENCE INTERVAL CALCULATIONS
Based on 1000 bootstrap replicates

```
CALL :
boot.ci(boot.out = resultsHW4_2F, type = "all")
```

Intervals :

Level	Normal	Basic	Studentized
95%	(25.31, 48.78)	(23.66, 47.92)	(27.46, 57.19)

Level	Percentile	BCa
95%	(26.57, 50.83)	(28.02, 53.82)

Calculations and Intervals on Original Scale
Some BCa intervals may be unstable