HW3

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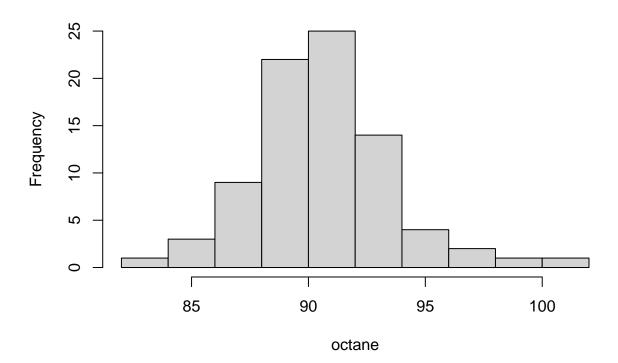
2023-06-09

```
octane = read.table("6-30.txt",header = TRUE)
octane=octane$Rating
o = sort(octane)
quantile(o, c(0.25,0.5,0.75),type=6)
##
      25%
             50%
                    75%
## 88.575 90.400 92.200
stem(octane, scale=2)
##
     The decimal point is at the |
##
##
##
      83 | 4
      84 | 33
##
      85 | 3
##
      86 | 777
##
      87 | 456789
##
      88 | 23334556679
##
##
      89 | 0233678899
##
      90 | 0111344456789
##
      91 | 0001112256688
      92 | 22236777
##
##
      93 | 023347
      94 | 2247
##
##
      95 |
      96 | 15
##
##
      97 |
      98 | 8
##
##
      99 |
##
     100 | 3
```

6-46 REL DENSity is 1/2????????

```
octane = read.table("6-30.txt",header = TRUE)
octane=octane$Rating
obj = hist(octane,breaks=8)
```

Histogram of octane



length(octane)

[1] 82

```
library(knitr)
library(kableExtra)
df = data.frame(Frequency = obj$counts)

n=length(obj$breaks)
for(i in 1:(n-2)){
   df$Class[i]=paste(obj$breaks[i],"$\\le x <$", obj$breaks[i+1])
}

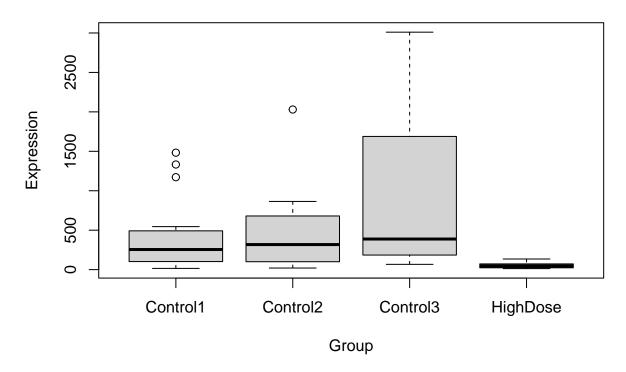
df$Class[n-1] = paste(obj$breaks[n-1],"$\\le x \\le$", obj$breaks[n])</pre>
```

Table 1: Freequency Distribution Table

Class	Frequency	Relative Frequency	Cumulative Frequency	Cumulative Relative Frequency
$82 \le x < 84$	1	0.0121951	1	0.0121951219512195
$84 \le x < 86$	3	0.0365854	4	0.0487804878048781
$86 \le x < 88$	9	0.1097561	13	0.158536585365854
$88 \le x < 90$	22	0.2682927	35	0.426829268292683
$90 \le x < 92$	25	0.3048780	60	0.731707317073171
$92 \le x < 94$	14	0.1707317	74	0.902439024390244
$94 \le x < 96$	4	0.0487805	78	0.951219512195122
$96 \le x < 98$	2	0.0243902	80	0.975609756097561
$98 \le x < 100$	1	0.0121951	81	0.98780487804878
$100 \le x \le 102$	1	0.0121951	82	1

```
df$"Relative Frequency"= obj$density*2
for(i in 1:(n-1)){
   df$"Cumulative Frequency"[i]=paste(cumsum(obj$counts)[i])
   df$"Cumulative Relative Frequency"[i]=paste(cumsum(obj$density*2)[i])
}
kable(df[c(2,1,3,4,5)], "latex", align="c", caption="Freequency Distribution Table", escape = FALSE)
```

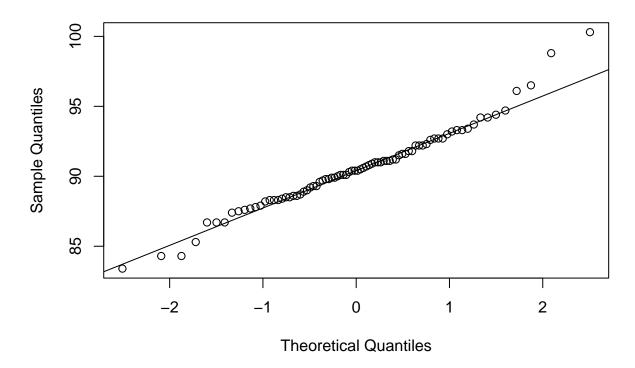
Boxplot of 4 Groups



The control groups seem to have the same median value but drastically different variances. The 'high dose' group has a mush smaller variance and slightly lower median expression as well.

```
octane = read.table("6-30.txt",header = TRUE)
octane=octane$Rating
qqnorm(octane)
qqline(octane)
```

Normal Q-Q Plot



The Normal Probability plot suggests that the data is normally distributed but seems to break down at both ends of the data range.

7-12

 $\mu_X=8.2\ minutes,\ n=49,\ \sigma_X=1.5\ minutes,\ \sigma_{\bar{x}}=\frac{\sigma_X}{\sqrt{n}}=0.2143$ Under Central Limit Theorem, \bar{X} is approx. normally distributed.

(a)

$$P(\bar{X} < 10) = P(Z < \frac{10-\mu}{\sigma_{\bar{X}}}) = P(Z < 8.4) = 1$$

(b)

$$P(5 < \bar{X} < 10) = P(\frac{5 - \mu}{\sigma_{\bar{X}}} < Z < \frac{10 - \mu}{\sigma_{\bar{X}}}) = P(Z < 8.4) - P(Z < -14.932) = 1 - 0 = 1$$

(c)

$$P(\bar{X} < 6) = P(Z < \frac{6-\mu}{\sigma_{\bar{X}}}) = P(Z < -10.27) = 0$$

7-37

a)

if $E(\hat{\Theta}) = \Theta$ then unbiased estimator $E(\hat{X}_1 - \hat{X}_2) = E(\hat{X}_1) - E(\hat{X}_2) = \mu_1 - \mu_2$. unbiased

b)

$$\sigma_{\bar{X}} = \sqrt{V(\hat{\Theta})} = \sqrt{V(\hat{X}_1 - \hat{X}_2)} = \sqrt{V(\hat{X}_1) + V(\hat{X}_2) + 2COV(\hat{X}_1, \hat{X}_2)} = \sqrt{\frac{\sigma_1^2}{n_1} + \frac{\sigma_2^2}{n_2}}$$

Estimating the standard error could be done by replacing the variance, σ with sample standard deviation, S.

c)

$$\begin{split} S_P^2 &= \frac{(n_1 - 1)S_1^2 + (n_2 - 1)S_2^2}{n_1 + n_2 - 2} \\ E(S_P^2) &= E(\frac{(n_1 - 1)S_1^2 + (n_2 - 1)S_2^2}{n_1 + n_2 - 2}) = \frac{1}{n_1 + n_2 - 2} \left((n_1 - 1)E(S_1^2) + (n_2 - 1)E(S_2^2) \right) = \frac{1}{n_1 + n_2 - 2} \left((n_1 - 1) * \sigma_1^2 + (n_2 - 1) * \sigma_2^2 \right) = \sigma^2 * \frac{n_1 + n_2 - 2}{n_1 + n_2 - 2} = \sigma^2 \end{split}$$

7-44 CHECK!!!!!!!!

$$f(x) = p(1-p)^{x-1}$$

$$L(p) = \prod_{i=1}^{n} p(1-p)^{x_i-1} = p^n (1-p)^{\sum_{i=1}^{n} x_i - n}$$

$$l(p) = nln(p) + (\sum_{i=1}^{n} x_i - n) ln(1-p)$$

$$\frac{\partial l}{\partial u} = 0 = \frac{n}{p} - \frac{\sum_{i=1}^{n} x_i - n}{1 - p} = \frac{n(1-p) - p \sum_{i=1}^{n} x_i - n}{p(1-p)} = n - p \sum_{i=1}^{n} x_i : p = \frac{n}{\sum_{i=1}^{n} x_i}$$

8-1

a)

$$z_0 = 2.14 \rightarrow P(Z < 2.14) = 0.9838 : \alpha = 2 \times (1 - 0.9838) \rightarrow CI = 100 \times (1 - \alpha) = 96.76\%$$

b)

$$z_0 = 2.49 \rightarrow P(Z < 2.49) = 0.9838 : \alpha = 2 \times (1 - 0.9963) \rightarrow CI = 100 \times (1 - \alpha) = 98.72\%$$

$$z_0 = 1.85 \rightarrow P(Z < 1.85) = 0.9678$$
 : $\alpha = 2 \times (1 - 0.9678) \rightarrow CI = 100 \times (1 - \alpha) = 93.56\%$

d)

$$z_0 = 2.00 \rightarrow P(Z < 2.00) = 0.9772$$
 $\therefore \alpha = (1 - 0.9772) \rightarrow CI = 100 \times (1 - \alpha) = 97.72\%$

e)

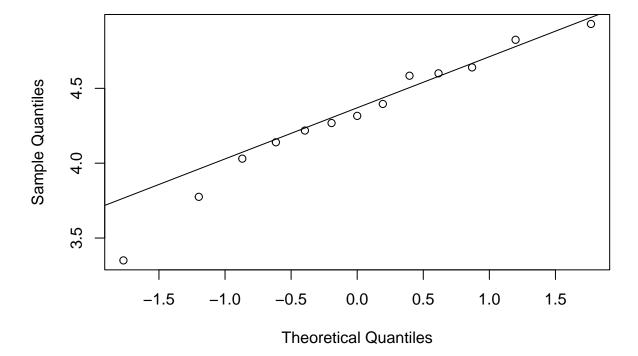
$$z_0 = 1.96 \rightarrow P(Z < 1.96) = 0.9750 : \alpha = (1 - 0.9750) \rightarrow CI = 100 \times (1 - \alpha) = 97.50\%$$

8-41

a)

speed = c(3.775302, 3.350679, 4.217981, 4.030324, 4.639692, 4.139665, 4.395575, 4.824257, 4.268119, 4.5
qqnorm(speed)
qqline(speed)

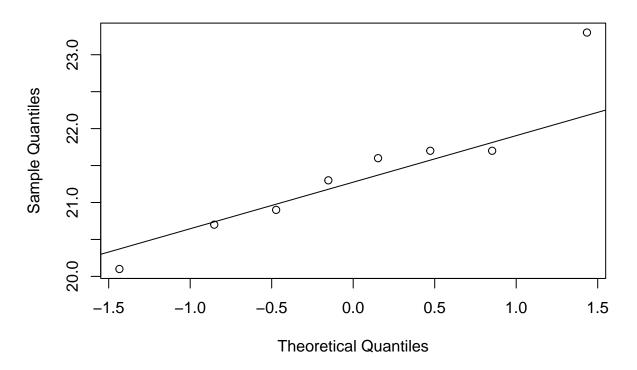
Normal Q-Q Plot



```
library(nortest)
ad.test(speed)
##
         Anderson-Darling normality test
##
##
## data: speed
## A = 0.23339, p-value = 0.7448
b)
 \bar{x} - t_{0.025,12} \left( \frac{s}{\sqrt{n}} \right) \le \mu \le \bar{x} + t_{0.025,12} \left( \frac{s}{\sqrt{n}} \right)   \mu = 4.313; \ s = 0.4328; \ n = 13; \ t_{0.025,12} = 2.179 
4.313 \pm 2.179 \left( \frac{0.4328}{\sqrt{13}} \right) \rightarrow 4.051 \le \mu \le 4.575
mean(speed)
## [1] 4.313222
sd(speed)
## [1] 0.4328017
length(speed)
## [1] 13
qt(.025,12)
## [1] -2.178813
qt(.05,12)
## [1] -1.782288
c)
\begin{split} \bar{x} - t_{0.05,12} \left( \frac{s}{\sqrt{n}} \right) &\leq \mu \\ \mu = 4.313; \ s = 0.4328; \ n = 13; \ t_{0.05,12} = 1.782 \\ 4.313 - 1.782 \left( \frac{0.4328}{\sqrt{13}} \right) &\rightarrow 4.099 \leq \mu \end{split}
```

```
temp = c(23.3, 21.7, 21.6, 21.7, 21.3, 20.7, 20.9, 20.1)
qqnorm(temp)
qqline(temp)
```

Normal Q-Q Plot



library(nortest)
ad.test(temp)

```
##
## Anderson-Darling normality test
##
## data: temp
## A = 0.35247, p-value = 0.3666
sd(temp)
```

[1] 0.9463275

length(temp)

[1] 8

 $\begin{array}{l} n=8; \ s=0.9463 \\ \chi^2_{\alpha/2,n-1}=\chi^2_{0.025,7}=16.012; \ \chi^2_{0.975,7}=1.689 \\ \frac{7(0.9463)^2}{16.012} \leq \sigma^2 \leq \frac{7(0.9463)^2}{1.689}=0.392 \leq \sigma^2 \leq 3.709 \rightarrow 0.626 \leq \sigma \leq 1.926 \end{array}$

 $\mathbf{Q2}$

a)

for
$$\mu_1$$
: $E(\frac{\sum_{n=1}^{X_i} \sum_{m=1}^{Y_j}}{2}) = \frac{1}{2} \left(E(\frac{\sum_{n=1}^{X_i}}{n}) + E(\frac{\sum_{m=1}^{X_j}}{m}) \right) = \frac{1}{2} \left(\frac{1}{n} E(\sum_{n=1}^{X_i} X_i) + \frac{1}{m} E(\sum_{n=1}^{X_j} X_i) \right) = \frac{1}{2} (\mu + \mu) = \mu$... unbiased

for
$$\mu_2$$
: $E(\frac{\sum X_i + \sum Y_j}{n+m}) = \frac{1}{n+m} (E(\sum X_i) + E(\sum Y_j)) = \frac{1}{n+m} (n\mu + m\mu) = \frac{n+m}{n+m} (\mu) = \mu$: unbiased

b)

for
$$\mu_1$$
: $Var(\frac{\sum_{n=1}^{X_i}\sum_{m=1}^{Y_j}}{2}) = \frac{1}{2}\left(\frac{1}{n^2}Var(\sum X_i) + \frac{1}{m^2}Var(\sum Y_j)\right) = \frac{1}{2}\left(\frac{1}{n^2}n\sigma^2 + \frac{1}{m^2}m\sigma^2\right) = \frac{1}{2}(\frac{\sigma^2}{n} + \frac{\sigma^2}{m}) = \frac{1}{2}(\frac{m\sigma^2}{nm} + \frac{n\sigma^2}{nm}) = \frac{\sigma^2}{2nm}(n+m)$
for μ_2 : $Var(\frac{\sum X_i + \sum Y_j}{n+m}) = \frac{1}{n+m}(Var(\sum X_i) + Var(\sum Y_j)) = \frac{1}{n+m}(n\sigma^2 + m\sigma^2) = \sigma^2$

c)

Relative Efficiency:
$$\frac{MSE(\hat{\mu}_1)}{MSE(\hat{\mu}_2)}$$

 $MSE(\hat{\mu}_1) = Var(\hat{\mu}_1) - Bias^2$: $Relative\ Efficiency = \frac{Var(\hat{\mu}_1)}{Var(\hat{\mu}_2)} = \frac{\sigma^2(n+m)}{\sigma^2} = \frac{n+m}{2nm}$

d)

Based on relative efficiency being less than 1, $\hat{\mu}_1$ is the better estimator.