# Home\_Work\_2

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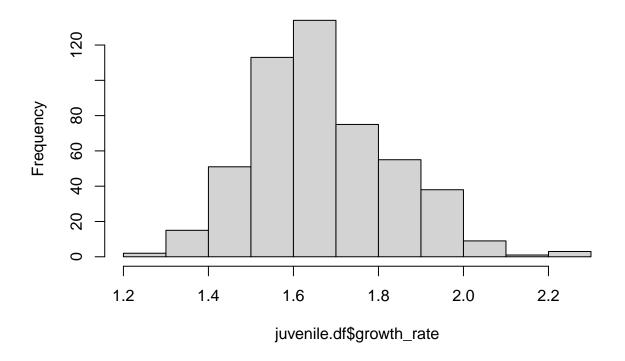
### ${\bf Question\_1}$

a). Plot a histogram of the variable growth\_rate

library(fishdata)

```
## Warning: package 'fishdata' was built under R version 4.0.5
juvenile.df <- juvenile_metrics
hist(juvenile.df$growth_rate)</pre>
```

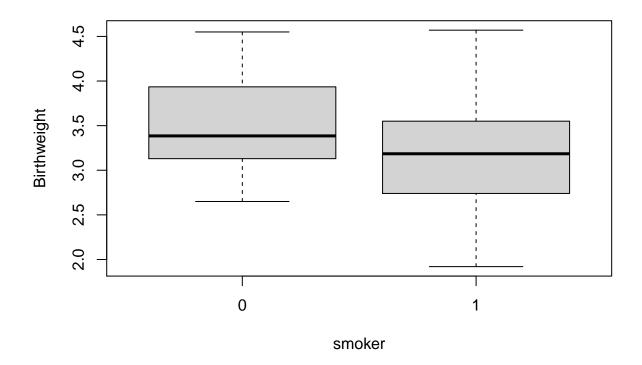
## Histogram of juvenile.df\$growth\_rate



Growth data doesn't seem to be normally distributed. i.e. data tends be skewed to slightly skewed.

b). Use the function ks.test in R to test a normality

```
sd(juvenile.df$growth_rate))
##
##
    One-sample Kolmogorov-Smirnov test
##
## data: juvenile.df$growth_rate
## D = 0.074733, p-value = 0.00785
## alternative hypothesis: two-sided
p-value = 0.00785, reject the null hypothesis that the growth rate data follows a normal distribution.
i.e. Accept the alternative hypothesis that growth rate data doesnot follow a normal distribution.
c).
growth.log <- log(juvenile.df$growth_rate)</pre>
t.test(growth.log, mu = 1.7)
##
##
    One Sample t-test
##
## data: growth.log
## t = -275.64, df = 495, p-value < 2.2e-16
## alternative hypothesis: true mean is not equal to 1.7
## 95 percent confidence interval:
## 0.4977464 0.5147647
## sample estimates:
## mean of x
## 0.5062556
exp(1.7-(mean(growth.log)))
## [1] 3.299413
growth rate is 3.2 times less than median growth rate.
Question_2
  a)
bw.df <- read.csv("Birthweight_reduced_kg_R.csv")</pre>
boxplot(Birthweight~smoker, data=bw.df)
```



T-test using calculated manaully

```
m1 <- length(bw.df$Birthweight[bw.df$smoker == 1])</pre>
m2 <- length(bw.df$Birthweight[bw.df$smoker == 0])</pre>
s2 <- var(bw.df$Birthweight[bw.df$smoker == 0])</pre>
s1 <- var(bw.df$Birthweight[bw.df$smoker == 1])</pre>
Sp2 < - (s1+s2)/2
test.stat <- (mean(bw.df$Birthweight[bw.df$smoker == 1])-</pre>
                 (mean(bw.df$Birthweight[bw.df$smoker == 0])))/sqrt(Sp2*(1/m1+1/m2))
test.stat
## [1] -2.103574
y \leftarrow ((m1+m2)-2)
2*(1-pt(abs(test.stat), y))
## [1] 0.04175107
T-test using R t.test function
t.test(bw.df$Birthweight[bw.df$smoker == 0], bw.df$Birthweight[bw.df$smoker == 1],
       conf.level = 0.90, alternative = "two.sided", var.equal = T)
##
##
    Two Sample t-test
## data: bw.df$Birthweight[bw.df$smoker == 0] and bw.df$Birthweight[bw.df$smoker == 1]
## t = 2.0934, df = 40, p-value = 0.0427
```

```
## alternative hypothesis: true difference in means is not equal to 0
## 90 percent confidence interval:
## 0.0734489 0.6773693
## sample estimates:
## mean of x mean of y
## 3.509500 3.134091
c).
mean_diff = mean(bw.df$Birthweight[bw.df$smoker == 0])- (mean(bw.df$Birthweight[bw.df$smoker == 1]))
s.e.m \leftarrow sqrt(Sp2*(1/m1+1/m2))
qtn \leftarrow abs(qt(0.05,((m1+m2)-2)))
Lower_cl <- mean_diff-(qtn*s.e.m)
Lower_cl
## [1] 0.07490483
Upper_cl <- mean_diff+(qtn*s.e.m)</pre>
Upper_cl
## [1] 0.6759134
manual computation: (0.074, 0.675) and using R-t.test function (0.0734, 0.677)
P < 0.05, suggests that there is moderate evidence for difference between mother's smoking habits.
We are 90% confident that the difference between mother's smoking habits is 0.0734 between 0.677.
```

#### Question\_3

```
chol.df <- read.csv("Cholesterol_R.csv")

a)

Experimental unit: 18 individuals diagnosed with high cholesterol who replaced butter

Population: individuals diagnosed with high cholesterol who replaced butter

Treatments: margarine [A and B]

Response: blood cholesterol levels

b)

boxplot(After8weeks~Margarine, data = chol.df)
```



C)

```
Computation of t.test by hand
```

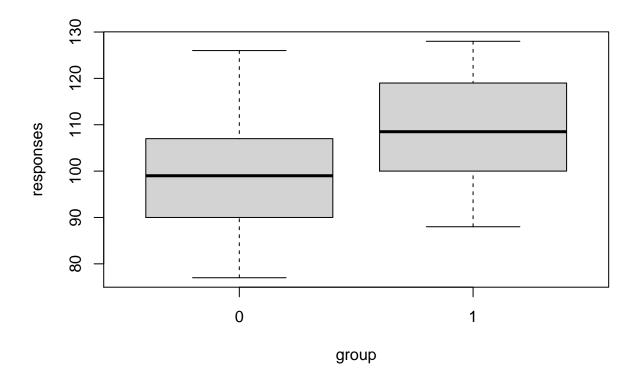
```
m1 <- length(chol.df$After8weeks[chol.df$Margarine == "A"])</pre>
m2 <- length(chol.df$After8weeks[chol.df$Margarine == "B"])</pre>
s2 <- var(chol.df$After8weeks[chol.df$Margarine == "A"])</pre>
s1 <- var(chol.df$After8weeks[chol.df$Margarine == "B"])</pre>
Sp2 \leftarrow (s1+s2)/2
test.stat <- (mean(chol.df$After8weeks[chol.df$Margarine == "A"])-</pre>
                  (mean(chol.df$After8weeks[chol.df$Margarine == "B"])))/sqrt(Sp2*(1/m1+1/m2))
test.stat
## [1] -1.125284
df \leftarrow ((m1+m2)-2)
2*(1-pt(abs(test.stat), df))
## [1] 0.2770664
T.test using R_t.test function
t.test(chol.df$After8weeks[chol.df$Margarine == "B"], chol.df$After8weeks[chol.df$Margarine == "A"],
       conf.level = 0.95, alternative = "two.sided", var.equal = T)
##
##
    Two Sample t-test
##
```

```
## data: chol.df$After8weeks[chol.df$Margarine == "B"] and chol.df$After8weeks[chol.df$Margarine == "A
## t = 1.1253, df = 16, p-value = 0.2771
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -0.5126535    1.6726535
## sample estimates:
## mean of x mean of y
## 6.068889    5.488889
    d)
(-0.5126535, 1.6726535)
    e)
```

P-value > 0.05, there is little or no evidence that there is difference in mean cholesterol reduction between the two brands of margarine after 8 weeks of use

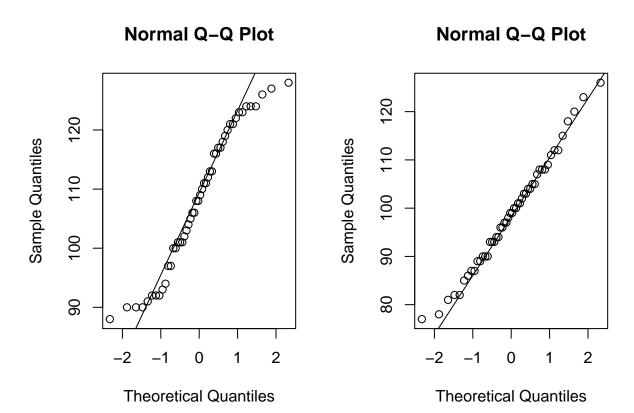
### ${\bf Question\_4}$

```
att.df <- read.csv("attendance.csv")
a)
boxplot(responses~group, data=att.df)</pre>
```



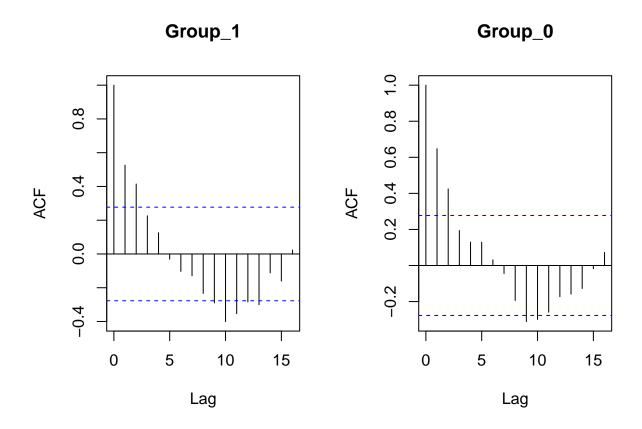
b)

```
par(mfrow=c(1,2))
qqnorm(att.df$responses[att.df$group == 1])
qqline(att.df$responses[att.df$group == 1])
qqnorm(att.df$responses[att.df$group == 0])
qqline(att.df$responses[att.df$group == 0])
```



Plots of the data look close to random samples from a normal distribution. seem normally distributed.

```
c).
par(mfrow=c(1,2))
acf(att.df$responses[att.df$group == 1], main="Group_1")
acf(att.df$responses[att.df$group == 0], main="Group_0")
```

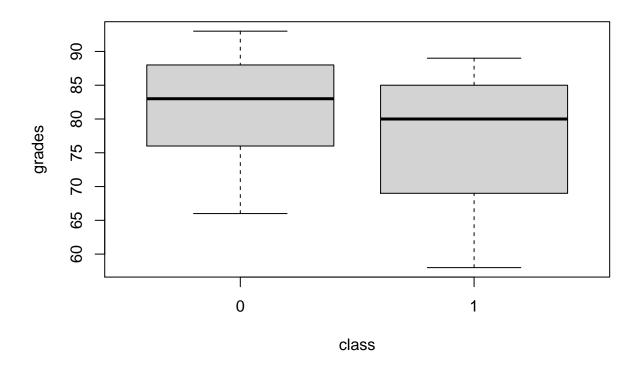


T.test wouldn't adequately account for the variation because there is correlation between points i.e. Assumption of independence is violated.

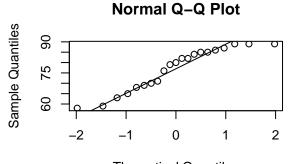
### ${\bf Question\_5}$

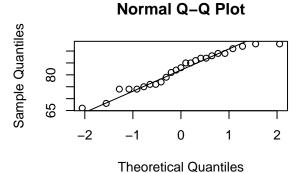
d)

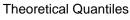
```
grad.df <- read.csv("grades.csv")
a)
boxplot(grades~class, data=grad.df)</pre>
```

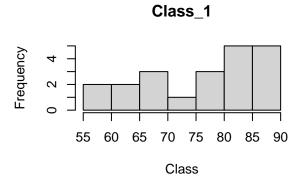


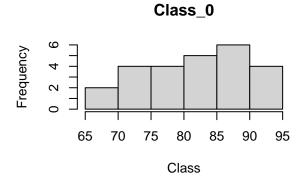
```
b)
par(mfrow=c(2,2))
qqnorm(grad.df$grades[grad.df$class== 1])
qqline(grad.df$grades[grad.df$class== 1])
qqnorm(grad.df$grades[grad.df$class== 0])
qqline(grad.df$grades[grad.df$class== 0])
hist(grad.df$grades[grad.df$class== 1], xlab="Class",main="Class_1")
hist(grad.df$grades[grad.df$class== 0], xlab="Class",main="Class_0")
```



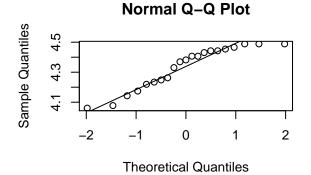


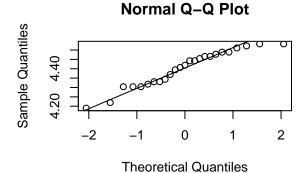


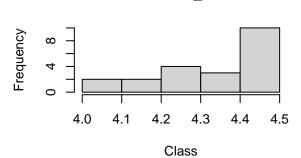




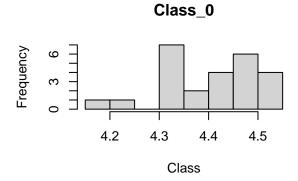
```
c)
grad.df$log_grades <- log(grad.df$grades)
par(mfrow=c(2,2))
qqnorm(grad.df$log_grades[grad.df$class== 1])
qqline(grad.df$log_grades[grad.df$class== 1])
qqnorm(grad.df$log_grades[grad.df$class== 0])
qqline(grad.df$log_grades[grad.df$class== 0])
hist(grad.df$log_grades[grad.df$class== 1], xlab="Class",main="Class_1")
hist(grad.df$log_grades[grad.df$class== 0], xlab="Class",main="Class_0")</pre>
```







Class\_1



c)

log transformation of grades didn't improve the normality of the grades.

d)

```
## grad.df$grades[grad.df$class == : cannot compute exact p-value with ties
##
## Wilcoxon rank sum test with continuity correction
##
## data: grad.df$grades[grad.df$class == 1] and grad.df$grades[grad.df$class == 0]
## W = 190.5, p-value = 0.1143
## alternative hypothesis: true location shift is not equal to 0
```