### Assessment01

### STAT414

#### 2024-12-08

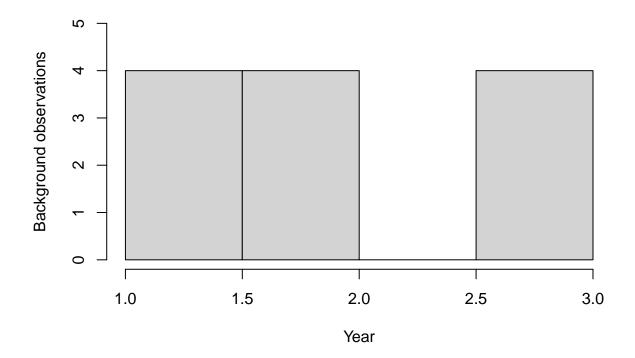
```
library(EnvStats)
##
## Attaching package: 'EnvStats'
## The following objects are masked from 'package:stats':
##
##
      predict, predict.lm
# 1. Arsenic concentrations (ppb) collected quarterly at two groundwater monitoring wells.
# These data are stored in the data frame EPA.92c.arsenic3.df.
data <- EPA.92c.arsenic3.df
background <- data[data$Well.type=="Background",]</pre>
compliance <- data[data$Well.type=="Compliance",]</pre>
background
     Arsenic Year Well.type
##
## 1
       12.6 1 Background
## 2
        30.8 1 Background
## 3
        52.0 1 Background
        28.1 1 Background
## 4
        33.3
## 5
                2 Background
        44.0 2 Background
## 6
## 7
        3.0 2 Background
        12.8 2 Background
## 8
## 9
        58.1 3 Background
## 10
        12.6 3 Background
## 11
        17.6 3 Background
        25.3
                3 Background
## 12
compliance
##
     Arsenic Year Well.type
## 13
        48.0 4 Compliance
## 14
        30.3 4 Compliance
## 15
        42.5 4 Compliance
## 16
        15.0
              4 Compliance
```

```
5 Compliance
## 18
          3.8
                 5 Compliance
                 5 Compliance
## 19
          2.6
## 20
         51.9
                 5 Compliance
# (a) For each well, plot the observations by year. Do you see any major differences
# between years?
hist(as.numeric(background$Year), xlab="Year",
     ylab="Background observations", ylim=c(0,5))
```

## 17

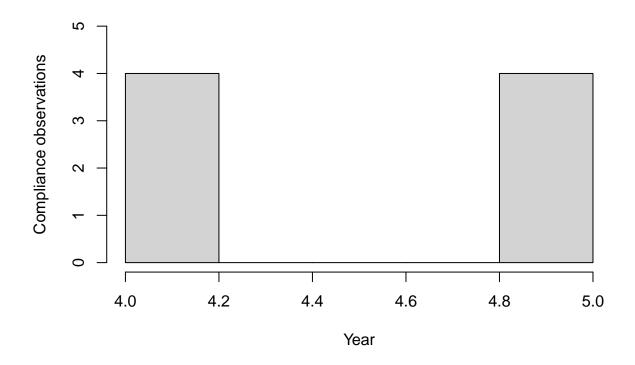
47.6

### Histogram of as.numeric(background\$Year)

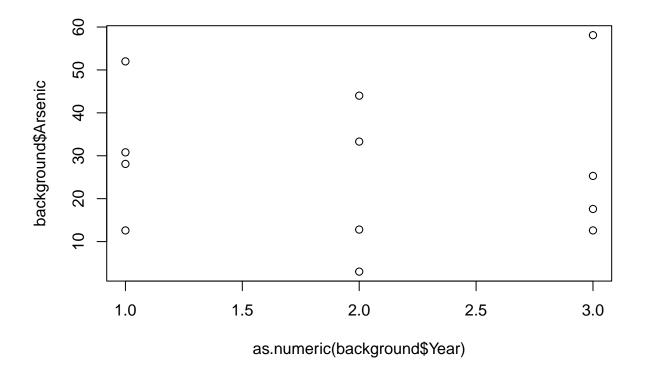


```
hist(as.numeric(compliance$Year), xlab="Year",
     ylab="Compliance observations", ylim=c(0,5))
```

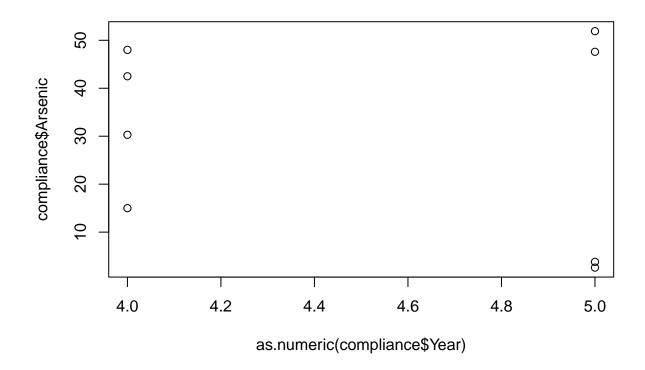
## Histogram of as.numeric(compliance\$Year)



plot(as.numeric(background\$Year), background\$Arsenic)



plot(as.numeric(compliance\$Year), compliance\$Arsenic)



cat("The background well data was collected years 1-3, and the compliance well
 was collected years 4-5. All data collected quarterly each year. No major
 differences except that the background well has 1 extra year of quarterly data
 collection, and the wells were not tested at the same time for any given year.")

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## was collected years 4-5. All data collected quarterly each year. No major
## differences except that the background well has 1 extra year of quarterly data
## collection, and the wells were not tested at the same time for any given year.
```

# # (b) Compute summary statistics for each well (combine years). summary(background)

```
##
       Arsenic
                     Year
                                 Well.type
##
   Min.
          : 3.00
                     1:4
                           Background:12
    1st Qu.:12.75
                     2:4
                           Compliance: 0
    Median :26.70
##
                     3:4
    Mean
           :27.52
                     4:0
##
    3rd Qu.:35.98
                     5:0
    Max.
           :58.10
summary(compliance)
```

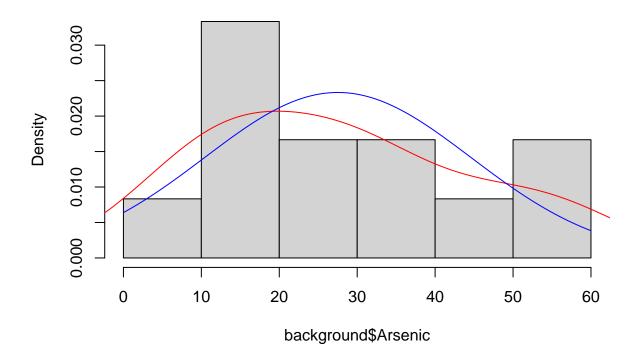
## Arsenic Year Well.type

```
Background:0
## Min. : 2.60
                   1:0
##
  1st Qu.:12.20
                   2:0
                         Compliance:8
                   3:0
## Median :36.40
          :30.21
                   4:4
## Mean
   3rd Qu.:47.70
                   5:4
##
  Max.
           :51.90
# (c) Compare the observed distribution of arsenic at each well. Use whatever types
# of plots you wish. Does the compliance well appear to show any evidence of
# contamination? Why or why not?
hist(background$Arsenic, probability="TRUE",main="Background arsenic")
lines(density(background$Arsenic), add=TRUE, col="red")
```

## Warning in plot.xy(xy.coords(x, y), type = type, ...): "add" is not a graphical
## parameter

```
#superimpose normal curve
curve(dnorm(x,mean=mean(background$Arsenic), sd=sd(background$Arsenic)), add=TRUE, col="blue")
```

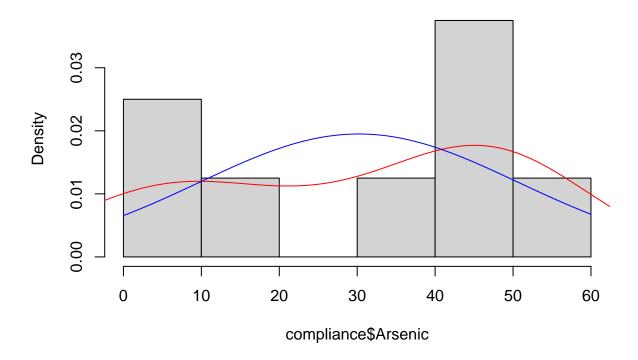
### **Background arsenic**



```
hist(compliance$Arsenic, probability="TRUE",main="Compliance arsenic")
lines(density(compliance$Arsenic), add=TRUE, col="red")
```

## Warning in plot.xy(xy.coords(x, y), type = type, ...): "add" is not a graphical
## parameter

### **Compliance arsenic**



cat("The compliance well does show evidence of aresenic contamination because its
 histogram distribution is slightly left-skewed, showing that it commonly has
 high ppb concentration of arsenic.")

## The compliance well does show evidence of aresenic contamination because its
## histogram distribution is slightly left-skewed, showing that it commonly has
## high ppb concentration of arsenic.

# 2. Consider only the data from the Background wells for this part.

# (a) Does it appear that the background well data may be modeled as coming from a # normal distribution? Support your conclusions with plots and tests of normality # statistics.

# qqplot(background\$Arsenic)
# qqline(background\$Arsenic)

cat("From the histogram we see that the shape of the distribution and its density
 line somehawt resembles the shape of the superimposed normal curve, therefore
 conclusions cannot be drawn from the distribution alone, there are noticable
 differences and errors in the shape while still remaining similar.")

```
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## line somehawt resembles the shape of the superimposed normal curve, therefore
## conclusions cannot be drawn from the distribution alone, there are noticable
## differences and errors in the shape while still remaining similar.
```

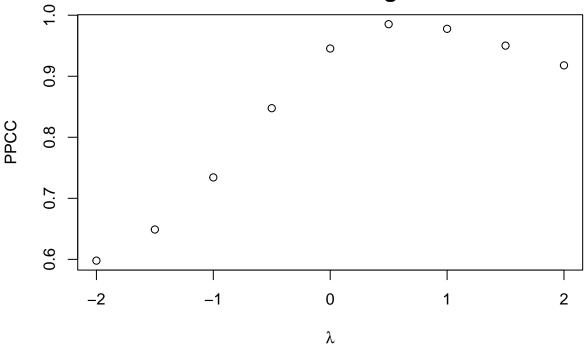
```
kurt <- kurtosis(background$Arsenic)
skew <- skewness(background$Arsenic)

cat("From the kurtosis and skewness values, we know that for a normal distribution
    their values should be 0 or close to 0. The value of skewness for the arsenic
    concentration for the background well is", skew, "and the kurtosis is", kurt,
    "
    neither of which are 0 or close to 0. Therefore we can say that from these values
    and the histogram distribution that the background well data is most likely not
    modeled from a normal distribution.")</pre>
```

## From the kurtosis and skewness values, we know that for a normal distribution
## their values should be 0 or close to 0. The value of skewness for the arsenic
concentration for the background well is 0.4895164 and the kurtosis is -0.6688492
## neither of which are 0 or close to 0. Therefore we can say that from these values
and the histogram distribution that the background well data is most likely not
## modeled from a normal distribution.

```
# (b) Find the value of lambda corresponding Box-Cox transformation of the data to normality.
#Plot a histogram of the transformed data and overlay the appropriate Normal curve to
#illustrate the quality of fit of this model. Repeat the above exercise with
#the original (non-transformed) data and illustrate how the transformation improves
#the normality fit.
box <- boxcox(background$Arsenic, lambda=c(-2,-1.5,-1,-.5,0,.5,1,1.5,2), plot="TRUE")
plot(box)</pre>
```

# **Box–Cox Transformation Results: PPCC vs. lambda for background\$Arsenic**



```
cat("Choose lambda = .5 since the PPCC value is maxmimized at lambda=.5")
```

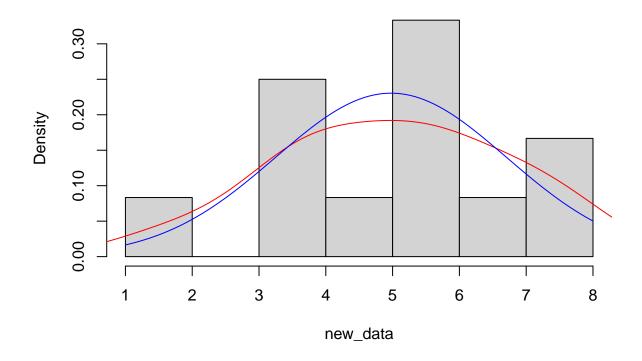
## Choose lambda = .5 since the PPCC value is maxmimized at lambda=.5

```
new_data <- background$Arsenic ^ .5
hist(new_data, probability="TRUE",main="Background lambda=.5 arsenic")
lines(density(new_data), add=TRUE, col="red")</pre>
```

## Warning in plot.xy(xy.coords(x, y), type = type, ...): "add" is not a graphical
## parameter

```
#superimpose normal curve
curve(dnorm(x,mean=mean(new_data), sd=sd(new_data)), add=TRUE, col="blue")
```

### Background lambda=.5 arsenic



cat("We can clearly see from the transformed data distribution that is matches the expected normal distribution, and that it is a much better fit compared to the original histogram distribution as calculated previously. The transformation improves the normality fit by making the density line similar in shape and structure to the superimposed normal curve.")

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## the expected normal distribution, and that it is a much better fit compared
## to the original histogram distribution as calculated previously. The transformation
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## to the superimposed normal curve.