Maternal Smoking and Infant deaths

Smoking by pregnant women may result in fetal injury, premature birth, and low birth weight.

- Is this warning to be taken seriously?
- Dataset: CHDS Berkeley, California.
- Taken entirely from Chapter 1 of the book
 Stat Labs: Mathematical Statistics Through Applications by
 Deborah Nolan and Terry P. Speed
- Check website: https://www.stat.berkeley.edu/users/statlabs/

Check: Normal Distribution

Do 68-95-99.7 first check to see if data is like normal or not.
 Compute

Skewness =
$$\frac{1}{n} \sum_{i=1}^{n} \left(\frac{x_i - \bar{x}}{\sigma_x} \right)^3$$
, Kurtosis = $\frac{1}{n} \sum_{i=1}^{n} \left(\frac{x_i - \bar{x}}{\sigma_x} \right)^4$.

- Skewness 0 indicates that the distribution is symmetric
- Kurtosis is a measure of the peak of the distribution.
- For standard normal Kurtosis is 3.

If Kurtosis and Skewness coefficients are *far* from 3 and 0 we can conclude that the data is not normal.

· Final check whether data is normal is to plot

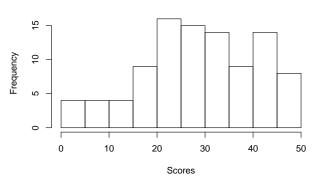
$$\left(z_{\frac{k}{n+1}},x_{(k)}\right)$$

• if the plot is a straight line then it indicates that data is normal

Are the Scores uniform?

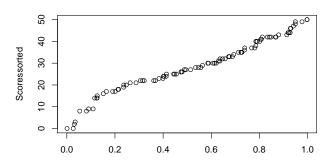
- > Scores = scan("Scores")
- > hist(Scores)

Histogram of Scores

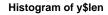


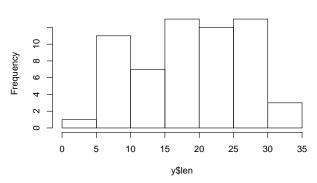
Can try the uniform-quantile plot

```
> Scores = scan("Scores")
> Scoressorted = sort(Scores)
> u = runif(97, 0,1)
> usorted = sort(u)
> plot(usorted, Scoressorted)
```

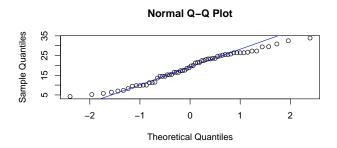


- > y = ToothGrowth
- > hist(y\$len)





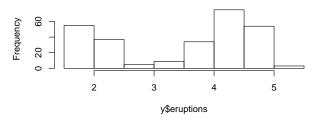
- > y = ToothGrowth
- > qqnorm(y\$len)
- > qqline(y\$len, col="blue") # adds a reference line



Notice that we did not scale or center. slope and intercept

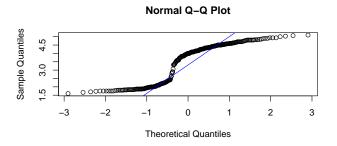
- > y = faithful
- > hist(y\$eruptions)

Histogram of y\$eruptions



Bi-Modal

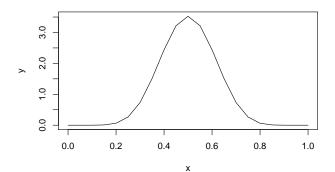
- > y = faithful
- > qqnorm(y\$eruptions)
- > qqline(y\$eruptions, col="blue") # adds a reference line



Notice that we did not scale or center. slope and intercept

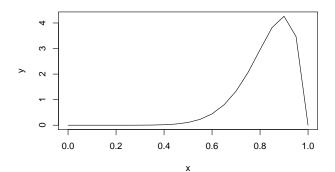
Beta-distribution

```
> x = seq(0,1, by=0.05)
> y = dbeta(x, 10,10)
> plot(x,y, type="l")
```



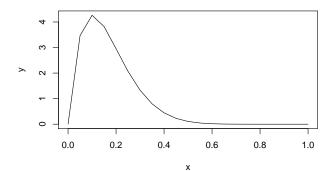
Beta-distribution

```
> x = seq(0,1, by=0.05)
> y = dbeta(x, 10,2)
> plot(x,y, type="l")
```



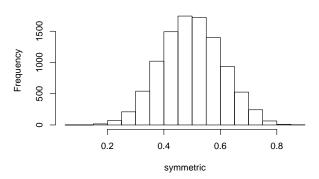
Beta-distribution

```
> x = seq(0,1, by=0.05)
> y = dbeta(x, 2,10)
> plot(x,y, type="l")
```



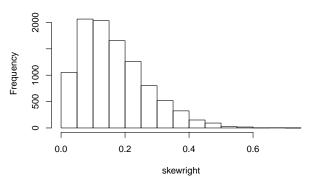
- > symmetric = rbeta(10000,10,10)
- > hist(symmetric)

Histogram of symmetric



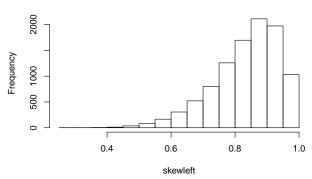
- > skewright= rbeta(10000,2,10)
- > hist(skewright)

Histogram of skewright

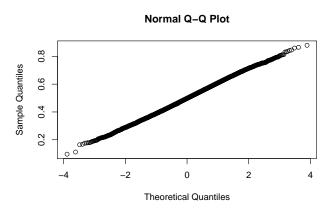


- > skewleft= rbeta(10000,10,2)
- > hist(skewleft)

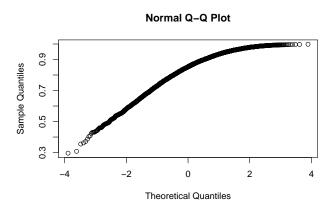
Histogram of skewleft



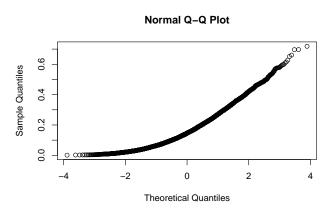
> qqnorm(symmetric)



> qqnorm(skewleft)



> qqnorm(skewright)



Skewness =
$$\frac{1}{n} \sum_{i=1}^{n} \left(\frac{x_i - \bar{x}}{\sigma_x} \right)^3$$
.

- Skewness is a measure of symmetry.
- Negative skewness will imply that the mean of the data is less than the median, and the data distribution is left-skewed.
- Positive skewness will imply that the mean of the data values is larger than the median, and the data distribution is right-skewed.

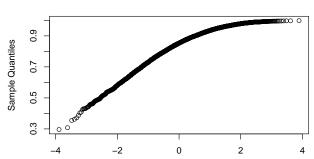
Skewness =
$$\frac{1}{n} \sum_{i=1}^{n} \left(\frac{x_i - \bar{x}}{\sigma_x} \right)^3$$
.

- In a histogram setting we can infer from mode:
 - Left Skew. Find the mode (the highest point of the distribution). The right of the mode should be shorter than the left of the mode.
 - Right Skew. Find the mode (the highest point of the distribution). The right of the mode should be longer than the left of the mode

Skewness =
$$\frac{1}{n} \sum_{i=1}^{n} \left(\frac{x_i - \bar{x}}{\sigma_x} \right)^3$$
.

- From Normal-Q plot
- > qqnorm(skewleft)

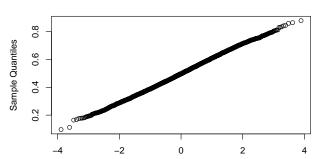
Normal Q-Q Plot



Skewness =
$$\frac{1}{n} \sum_{i=1}^{n} \left(\frac{x_i - \bar{x}}{\sigma_x} \right)^3$$
.

- From Normal-Q plot
- > qqnorm(symmetric)

Normal Q-Q Plot



Skewness =
$$\frac{1}{n} \sum_{i=1}^{n} \left(\frac{x_i - \bar{x}}{\sigma_x} \right)^3$$
.

- From Normal-Q plot
- > qqnorm(skewright)

Normal Q-Q Plot

