Johnson SU distribution

MLE (Maximum Likelihood Estimate) fit to determine parameters LR (Likelihood Ratio) approach to find tolerance limit

Pseudo-Code Used to Describe Process

Input

```
x <- iris$Sepal.Width
sided <- 1
alpha <- 0.01 # confidence = 1 - alpha / sided
P <- 0.99 # proportion or coverage</pre>
```

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3.0
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101
```

Determine Johnson Su Parameters

```
## Define nll (negative log likelihood) function to fit
## parameters: gamma, delta, xi, lambda
nll <- function() {
   pdf <- delta /( lambda * sqrt(2 * pi) ) *
        1 / sqrt(1 + (x-xi)/lambda )^2) *
        exp( -0.5*(gamma + delta * asinh((x-xi)/lambda ))^2 )
   nll <- -sum(log(pdf))
}</pre>
```

MLE fit to minimize nnl() returns gamma, delta, xi, and lambda

```
gamma delta xi lambda quant
standard fit -3.306484 5.319412 1.784619 1.887725 NA
```

Calculate Tolerance Limit

(1) Determine Equivalent Johnson SU Fit Using Quantile as a Parameter Instead of Gamma

MLE fit to minimize nll.q() returns quant, delta, xi, and lambda for given P

```
gamma delta xi lambda quant
standard fit -3.306484 5.319412 1.784619 1.887725 NA
fit on quantile at 1-P -3.306500 5.319415 1.784613 1.887714 2.134413
fit on quantile at P -3.307633 5.319502 1.784351 1.887396 4.178574
```

(2) Calculate Confidence Limits on Quantile Using LR (Likelihood Ratio)

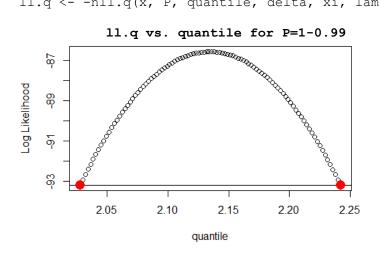
Find peak (ll.max) of log likelihood function (-nll.q())

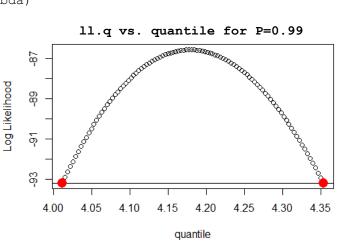
11.max = -86.55915

Reduce peak by chi-squared
11.tol <- 11.max - qchisq(1 - alpha/sided, 1) # qchisq(1-0.01/1, 1) = 6.634897</pre>

11.to1 = -93.19405

Confidence limits are the intersection of ll.tol and the log likelihood function ## for given level of coverage, P ll.q <--nll.q(x, P, quantile, delta, xi, lambda)



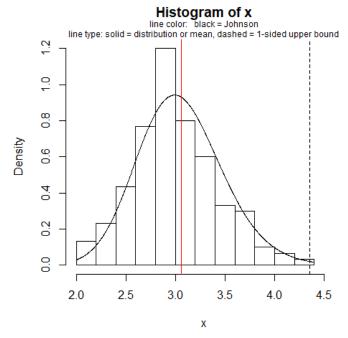


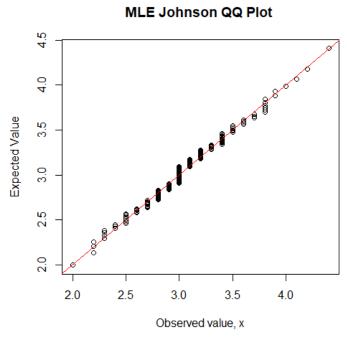
Final confidence interval for P= 0.01 2.027558 2.242365

Final confidence interval for P= 0.99 4.011839 4.352955

Based on the above:

Histogram with Upper, 1-Sided, 99/99 Tolerance Limit and QQ Plot





Actual Coding Used to Recreate Above Output

Acknowledgement

The approach described in the pseudo-code section is based on the approach found here: $https://personal.psu.edu/abs12/stat504/Lecture/lec3_4up.pdf$