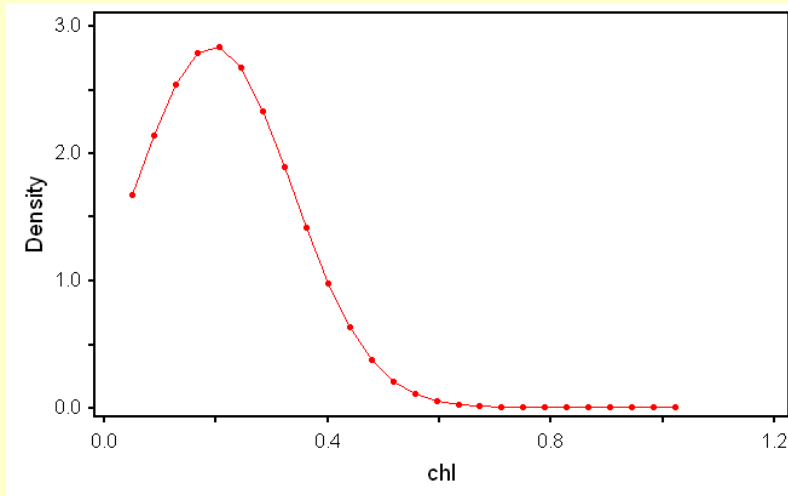
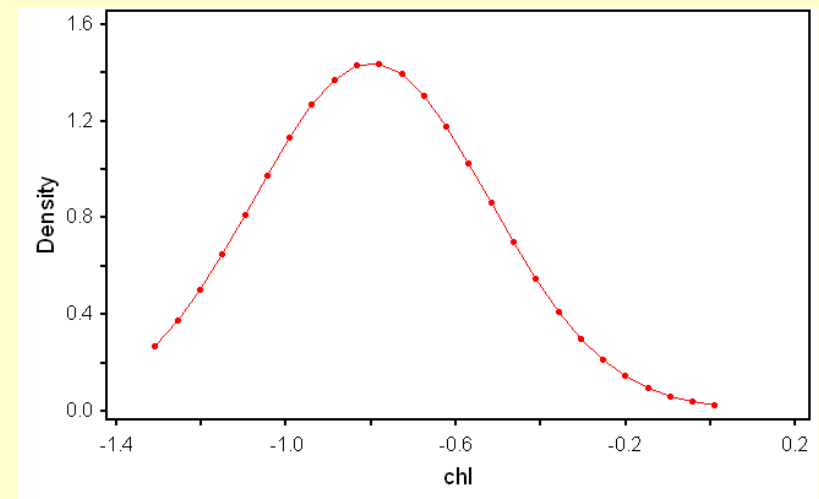


Data Distributions and Transformations



Chl Concentration



Log (Chl Conc.)

Approach

- We can transform the data to achieve normality
- Need to implement monotonic transformations:

Actual Values Change

Ranks Do Not Change

Presence / Absence Transformation

NONMONOTONIC TRANSFORMATIONS

Reasonable and acceptable
domain of x

Range of $f(x)$

x^0 (power)

all

0 or 1 only

P / A

(x)  **f(x)**

Note: 0 power transformation is NOT monotonic

It recodes data as Presence / Absence (0 / 1)

Log Transformation

Logarithmic transformation $f(x) = \ln(x)$ OR $\log(x)$

TRANSFORMATION	Reasonable and acceptable domain of x	Range of $f(x)$
$\log(x)$	positive	all

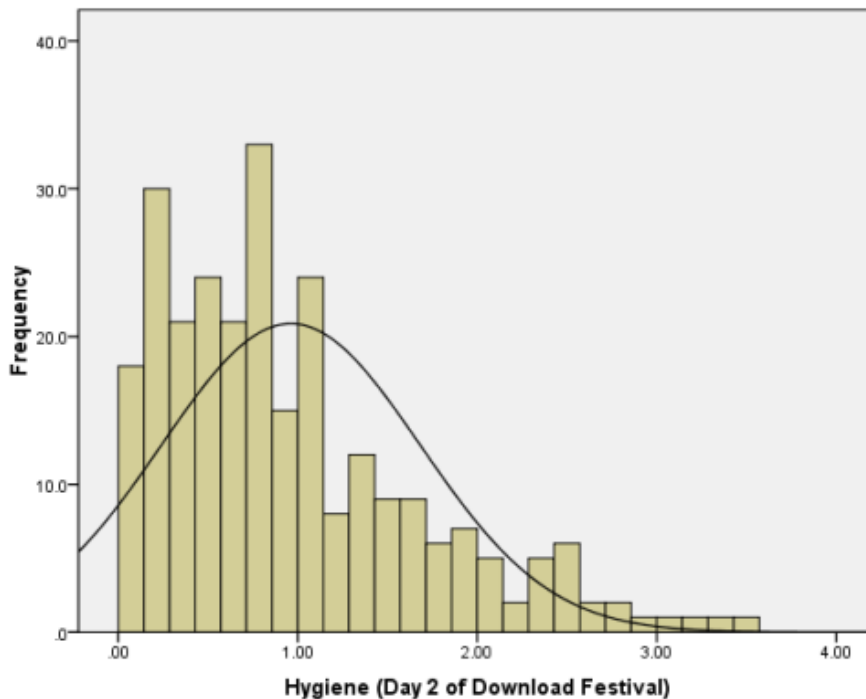
(x)  $f(x)$

- This transformation is useful when:
 - High degree of variation within samples (e.g., Chl Conc.)
 - Large outliers (tails) and lots of zeros
- Note: to log-transform data containing zeros, a small number should be added to all data points.
 - With count data, add one, so that: $f(x) = \log(0+1) = 0$
 - With density data, add constant smaller than smallest possible sample, so that: $f(x) = \log(0+0.001) = -3$

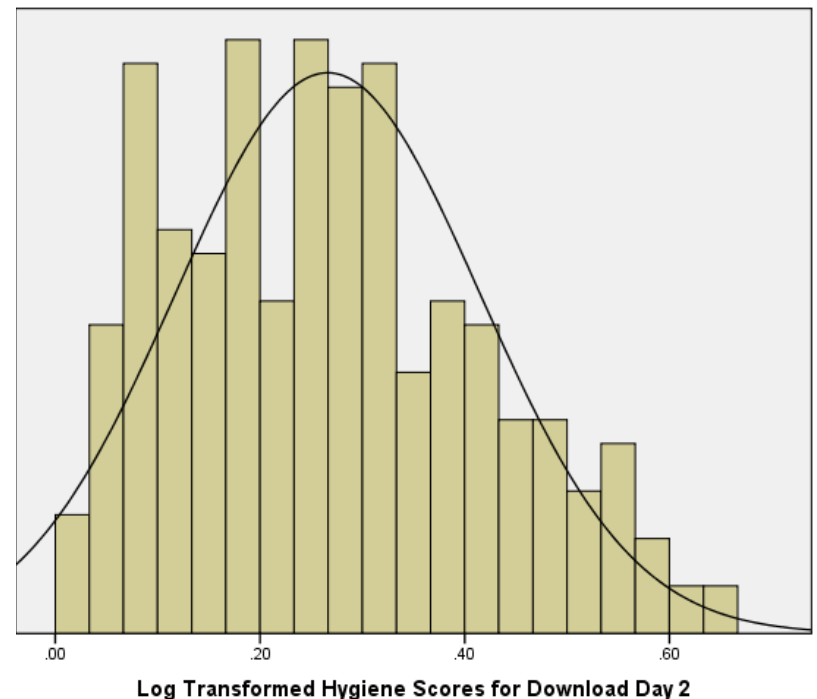
Log Transformation

Log Transformation ($\log(X_i)$): Reduces positive skew

Before



After



Square Root Transformations

MONOTONIC TRANSFORMATIONS

Reasonable and acceptable domain of x

Range of $f(x)$

$x^{1/2}$ (power)

nonnegative

nonnegative

(x)



f(x)

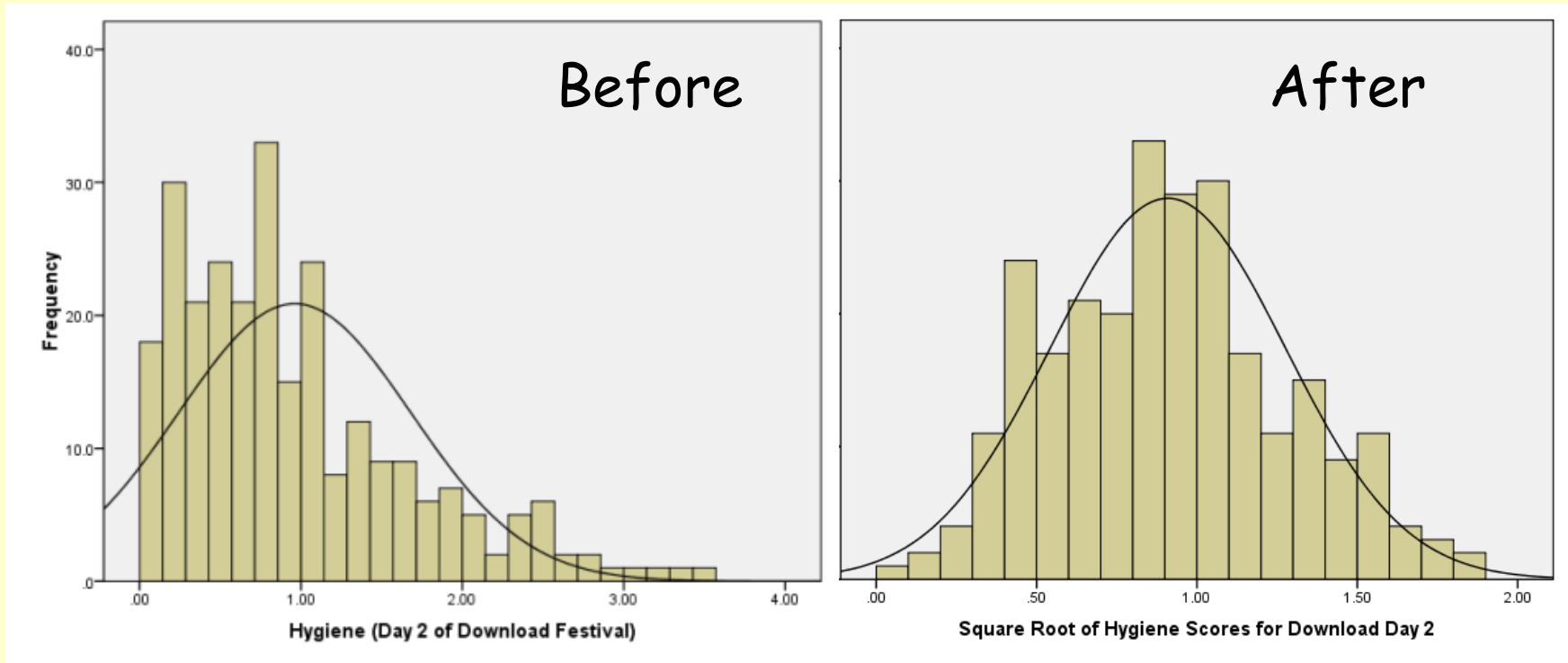
Power exponents:

$\frac{1}{2}$ power (square root)

Square root transform deals with positive skew, by bringing in large tails.
Special treatment of zeros not necessary.

Square Root Transformation

Square Root Transformation ($\sqrt{X_i}$):
Reduces positive skew. Useful for stabilizing variance



Data Transformations - For Proportions

Arcsine / Arcsine-squareroot transformation

TRANSFORMATION	Reasonable and acceptable domain of x	Range of $f(x)$
$\arcsin(x)$	$0 \leq x \leq 1$	0 to 1 inclusive
$\arcsin(x^{1/2})$	$0 \leq x \leq 1$	0 to 1 inclusive

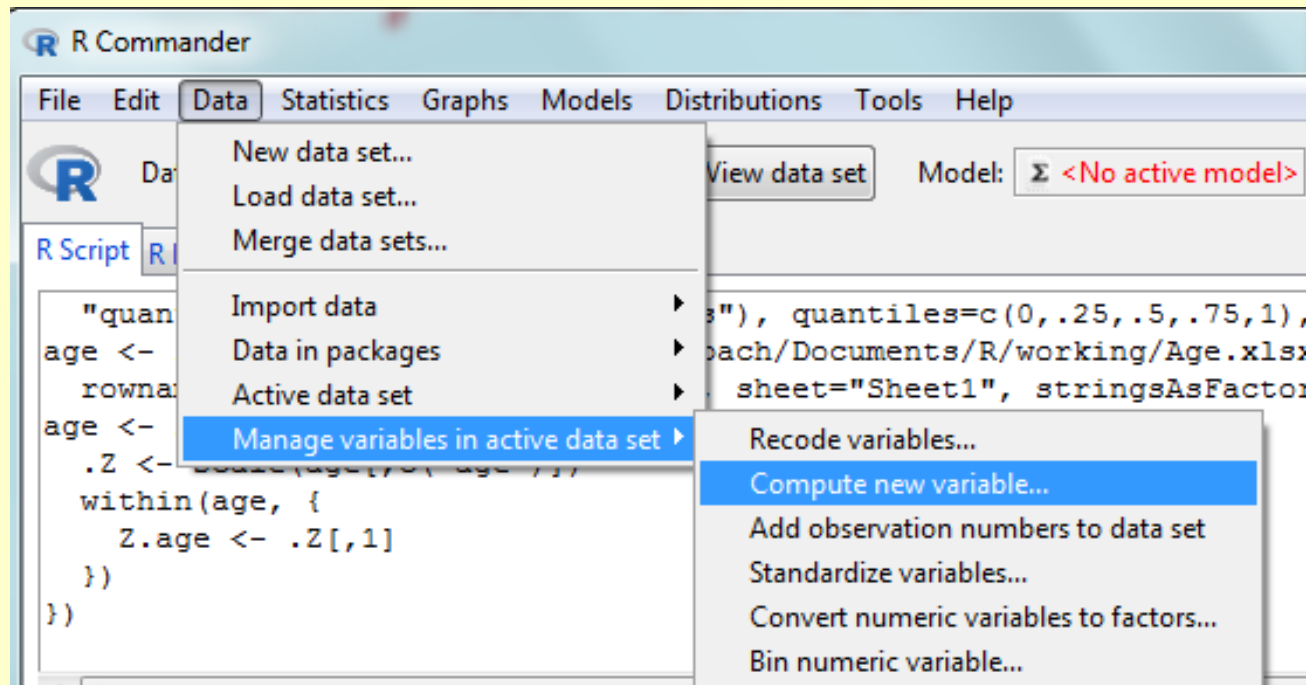
➤ This transformation is useful when dealing with proportional data (e.g., Percent Cover)

➤ Note: data must range between 0 and 1, inclusive.

The constant $2 / \pi$ scales the result of $\arcsin(x)$ [in radians] to range from 0 to 1, assuming that $0 \leq x \leq 1$.

So, the Data
are not Normal...
Now What ?

Transform the
Data in Rcmdr to
Fix the Problem

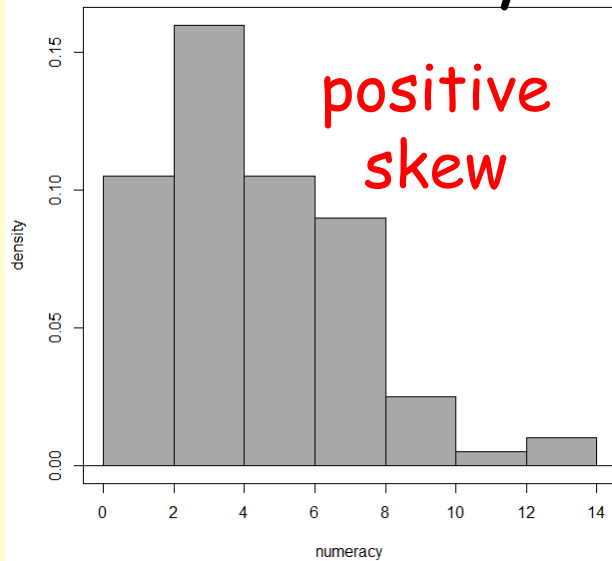


Data

Manage variables in active data set

Compute New Variable

Numeracy:

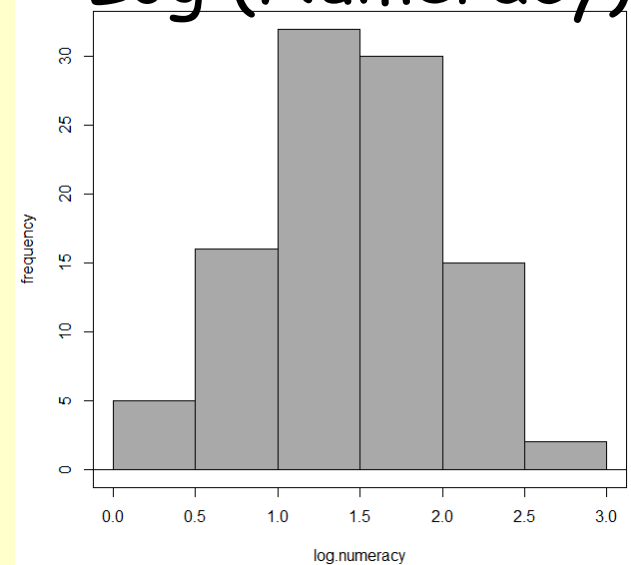


> Summary

Min: 1

Max = 14

Log (Numeracy):



Compute New Variable

Current variables (double-click to expression)

computer
exam
lectures
numeracy
uni
variable

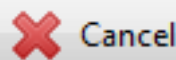
NOTE: In R:
 $\text{Log} = \text{Ln}$
 $\text{Log}_{10} = \text{Log}$

New variable name

log.numeracy

Expression to compute

log (numeracy)



log.numeracy

1.9459101
0.0000000
1.7917595
1.7917595
0.6931472
2.0794415
1.0986123
1.9459101
1.3862944
1.7917595
1.0986123
1.0986123
1.0986123
0.0000000

Hints for Computing New Variables

Log = is the natural ln

Log10 = is the log base 10

asin = arcsine

sqrt = square root

Summary

Rules for Data Transformations

Most Important Rule: Do not Reverse the Order of the Values (larger remains larger... smaller remains smaller)

Monotonic: change values but retain ranks

Non-monotonic: change values and ranks

(For example: Add random number, Multiply by (-1))

Summary

Take-home Lessons

- Parametric tests are more powerful, but are based on assumption of normally distributed data
- Determine normality criteria and undertake data transformations, if needed
- If you are unsure, data transformations can always be attempted to compare the same test results, using transformed and un-transformed data
- Test normality before / after data transformations
- If transformations do not work...
use non-parametric tests