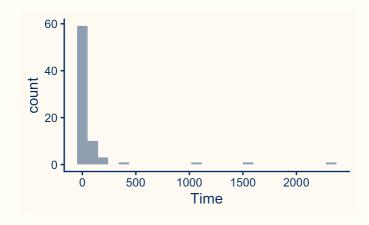
# Remedial Measures: Transformations

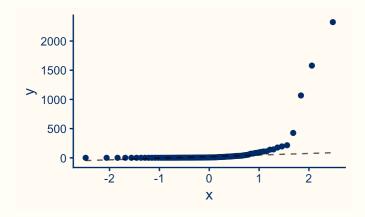
Stat 230: Applied Regression Analysis

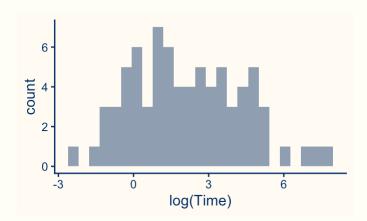
## PDF version of slides

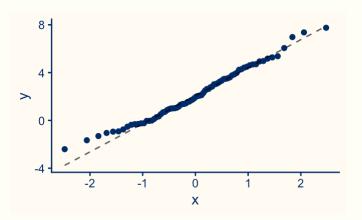
## Easing skew

If a set of data values is skewed to the right, taking the (natural) log of each data value *can* result in a data set that is roughly symmetric and often roughly normal.

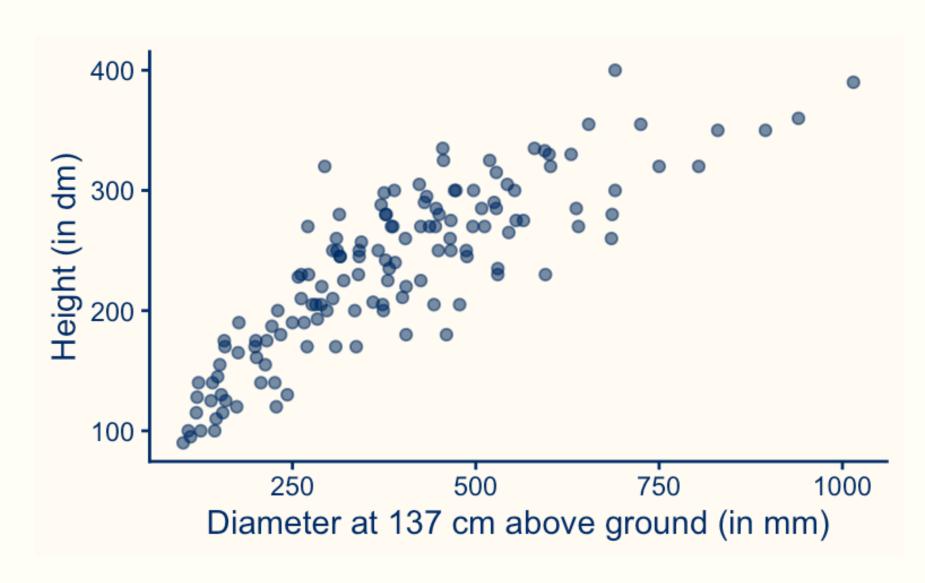




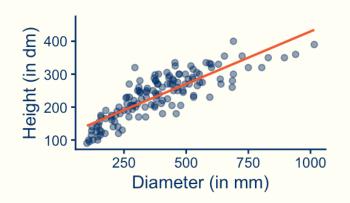


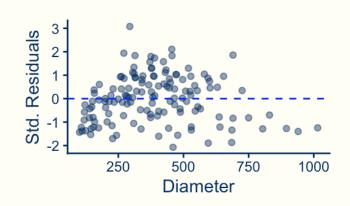


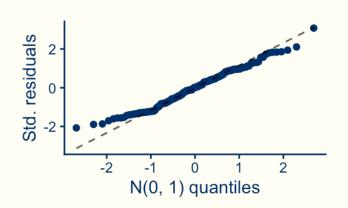
## How are tree height and tree diameter related for the western red cedar?



#### Are the conditions violated?



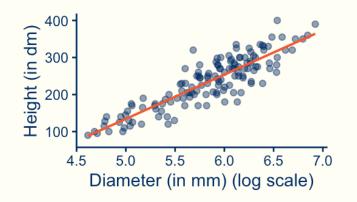


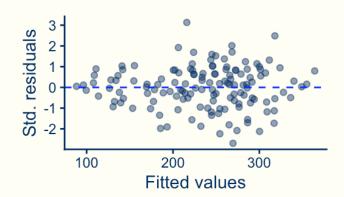


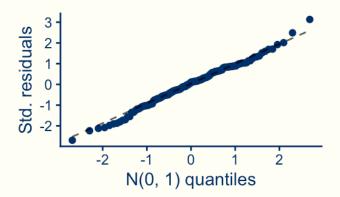
- a. linearity
- b. constant errors
- c. independent errors

- d. normal errors
- e. outliers
- f. none

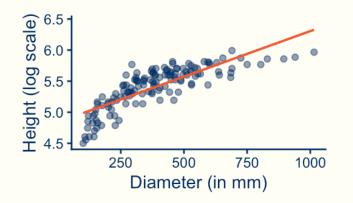
## Does transforming X help?

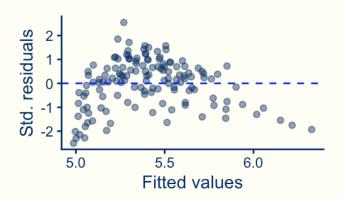


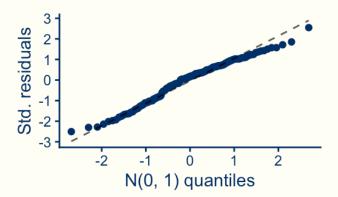




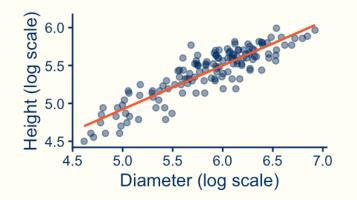
## Does transforming Y help?

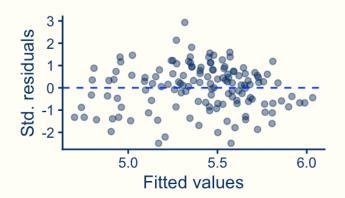


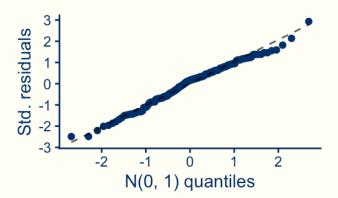




## Transforming both X and Y?







#### Your turn

- Work through the first example on the handout with your neighbor(s)
- Online version with R chunks:

# Back-Transforming

Converting a transformed variable back to its original scale

## **Back-Transforming**

Log scale

Original scale

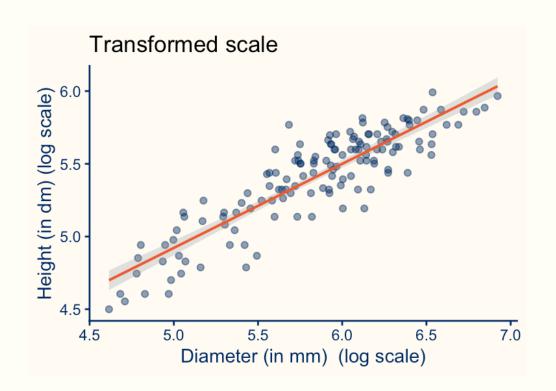
mean	mean
2.146	98.558

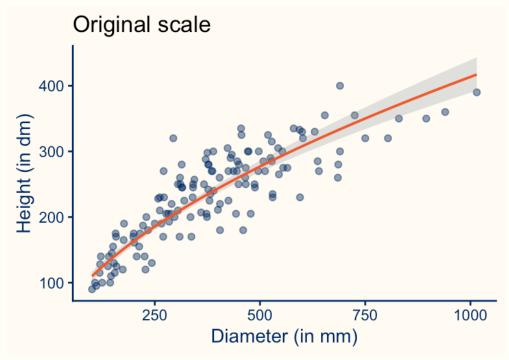
• Back-transformed mean:  $e^{2.146} \approx 8.55$ 

#### i R Note

- log is the natural log
- $\exp(x)$  calculated  $e^x$

## Displaying a transformed model

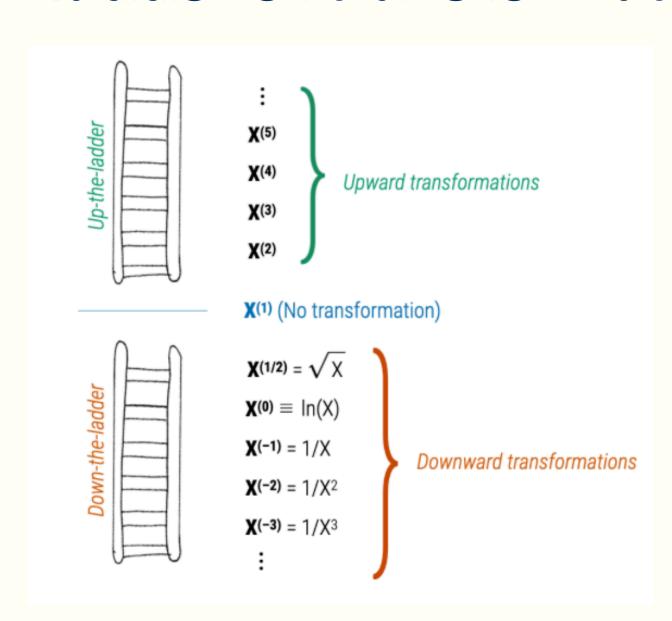




#### Rules of thumb

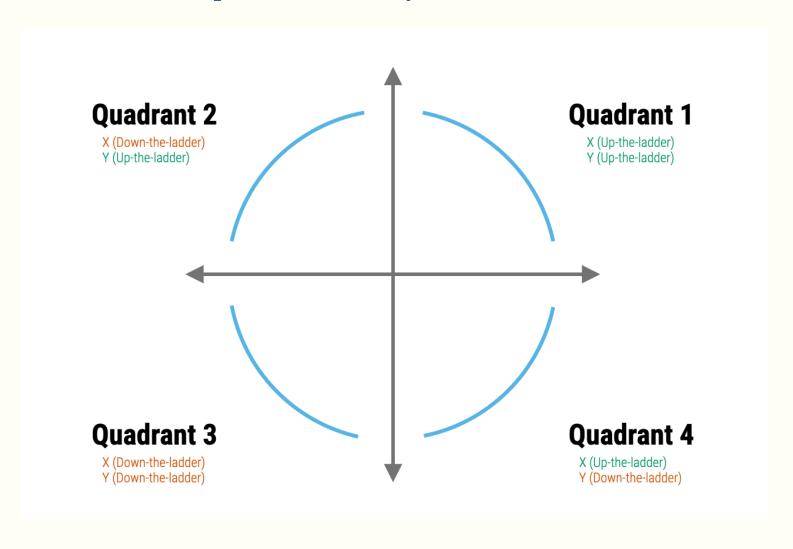
- **transform x**: if mean function is nonlinear, but is monotonic and the residual variance is constant
- **transform y**: if mean function is nonlinear and the residual variance increases as the mean increases (log, reciprocal, or square root often work)
- **log rule**: if values range over more than 1 order of magnitude and are strictly positive, then the natural log is likely helpful
- range rule: if the range is considerably less than 1 order of magnitude, then transformations are unlikely to help
- square roots are useful for count data

#### Ladder of transformations



## Rule of the Bulge

Introduced by John Tukey and Frederick Mosteller for "straightening" data to better meet the assumption of linearity



# Interpreting a logtransformed model

## **Back-Transforming**

Log scale

Original scale

mean	median	mean	median
2.146	1.933	98.558	6.925

- Back-transformed median:  $e^{1.933} \approx 6.91$
- Back-transformed mean:  $e^{2.146} \approx 8.55$

## **Back-transforming log transformations**

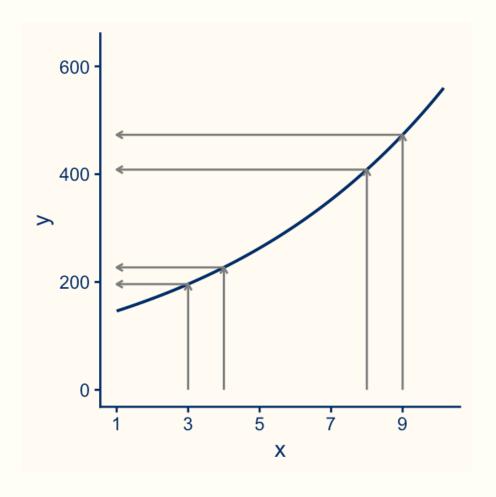
- Often log-transforming a variable makes in approximately symmetric
- If symmetric, then the median ≈ mean on the log scale

• Inference made mean on the log scale can thought of as inference for the median on the log scale

## Log-transform of Y only

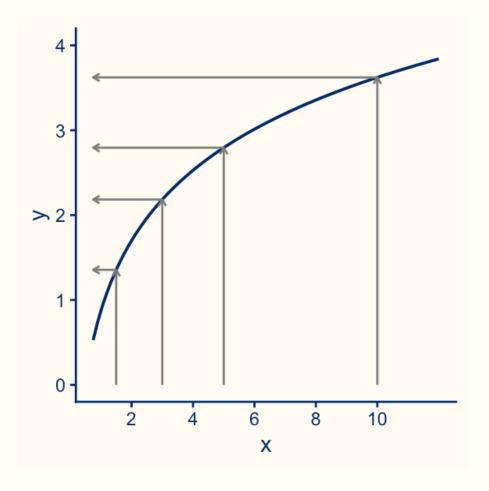
The median of at is times larger (smaller) than the median of at .

Or... increasing by 1 increases (decreases) the median of by a factor of .



## Log-transform of X only

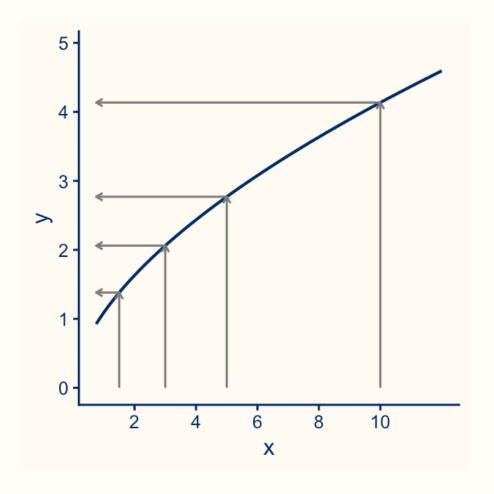
A doubling of is associated with the mean response increasing (decreasing) by units.



## Log-transform both Y and X

The median of at is times greater (smaller) than the median of at .

Or... A doubling of x is associated with the median of Y increasing (decreasing) by a factor of .



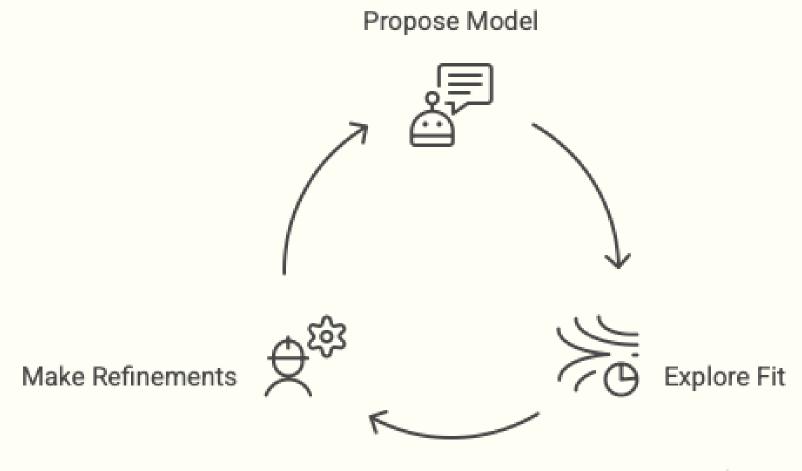
#### Your turn

You estimated the model

term	estimate s	std.error statistic p.value
(Intercept)	2.190	0.176 12.439 < 0.001
log(bodyweight)	0.759	0.042 18.163 < 0.001

- 1. Interpret the slope in context
- 2. Interpret the intercept in context

## Modeling is an iterative process



Made with ≽ Napkin

#### Issues with transformations

- You're often guessing Statistics is an art AND a science!
- Changes the interpretation of the parameters need to back-transform to provide interpretable results
- Changes SEs of the parameters
- Not always easy to keep track of all your assumptions