

# Generalized Linear Models

## Violating Linear Model Assumptions

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### Fitting a linear model

#### Linear model

$$y_i = \beta_0 + \beta_1 X_1 + \epsilon_i$$

$$\epsilon_i \sim \mathcal{N}(\mu, \sigma)$$

Minimize residuals

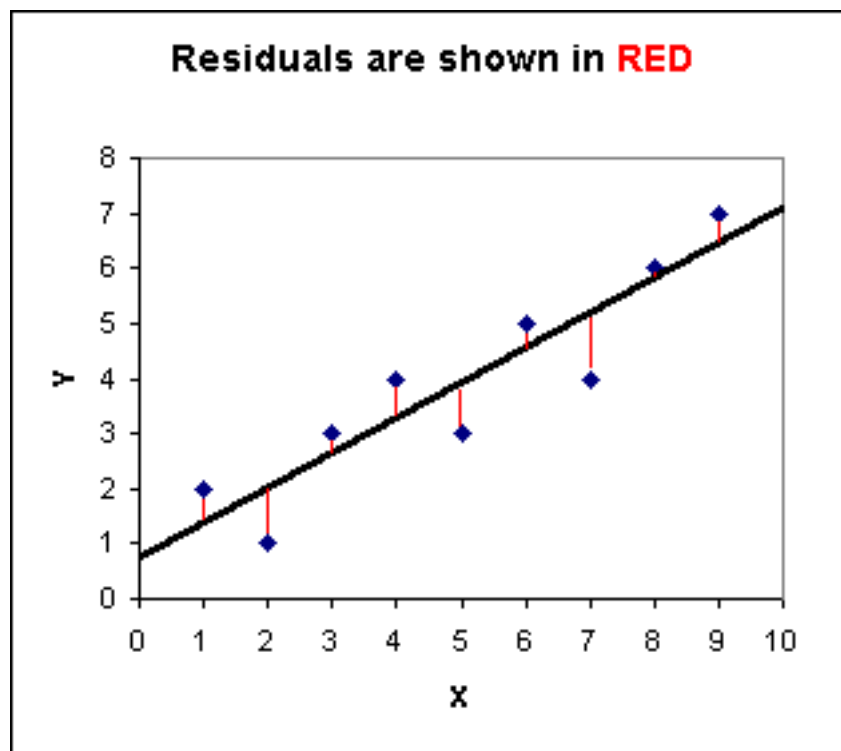


Figure 1: Residuals

<http://mtweb.mtsu.edu/stats/dictionary/formula.htm>

### Review: Model assumptions

What are our linear model assumptions?

### Review: Model assumptions

- Linearity
- Homogeneity of variance (Homoscedasticity)

- Normally distributed error
- Minimal multicollinearity (if multiple  $X$ )
- Independence of observations (no autocorrelation)

## Violating assumptions in ecology

In population biology and ecology, violating these rules is more of a rule than an exception.

What we do depends on the assumption.

## Normality

Possibly of least concern. Given a large sample size, models are often robust to modest violations of normality.

However, we don't always have large sample sizes in ecological data sets.

## Other violations

- Linearity - will likely result in issues of homogeneity. Means your model doesn't fit and you should develop a different deterministic model
- Homogeneity - small variation is okay but otherwise a major problem. Can transform the data or make different distributional assumptions
- Multicollinearity - only a problem at high values, partially an estimation problem, partly a problem of inference. Can also bias parameter uncertainty
- Independence - The most serious problem. Can alter the model (model the residuals), transform, or group the data to address these issues.

## Case Study 1: Mallard Data

```
ducks <- read.table("Data/mallard_counts.csv", sep = ",", header = TRUE, stringsAsFactors = FALSE)
```

Counts

Log transform

Analyze

Plot

Better solution is GLM

Poisson distribution

## Generalized Linear Model

Example with a Poisson distribution for count data. **Why do we use Poisson for counts?**

1. Distribution:  $C_i \sim \text{Poisson}(\lambda_i)$
2. Link Function:  $\log(\lambda_i)$
3. Linear Predictor:  $\log(\lambda_i) = \alpha + \beta X_i$

Example when data is 0/1

When is this the case?

## Case Study 2: Wedge Clam Data

Biomass of 398 wedge clams. Explore the data with a scatterplot matrix.