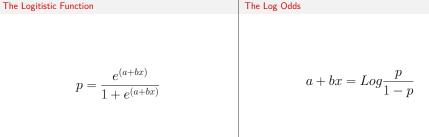
Logistic Regression	Allegations of the second of t
The Logitistic Function	The Log Odds

The Logitistic Curve for Probabilities



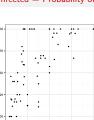
The Odds

 $Odds = \frac{p}{1-p}$ 

Cryptosporidium



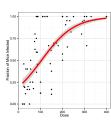
Fraction of Mice Infected = Probability of Infection



## Two Different Ways of Writing the Model

```
# 1) using Heads, Tails
glm(cbind(Y, N-Y) ~ Dose, data=crypto, family=binomial)
# 2) using weights as size parameter for Binomial
glm(Y/N ~ Dose, weights=N, data=crypto, family=binomial)
```

### The Fit Model



## The Fit Model

```
# Call:
 glm(formula = cbind(Y, N - Y) ~ Dose, family = binomial, data = crypt
 Deviance Residuals:
             1Q Median
 -3.953 -1.244 0.233
                        1 553
                                 3,601
# Coefficients:
             Estimate Std. Error z value Pr(>|z|)
# (Intercept) -1.40777
                         0.14848
                                  -9.48 <2e-16
# Dose
              0.01347 0.00105 12.87 <2e-16
# (Dispersion parameter for binomial family taken to be 1)
     Null deviance: 434.34 on 67 degrees of freedom
# Residual deviance: 200.51 on 66 degrees of freedom
# AIC: 327
```

#### The Meaning of a Logit Coefficient

Logit Coefficient: A 1 unit increase in a predictor = an increase of  $\beta$  increase in the log-odds ratio of the response.

$$Log\ Odds\ Ratio = Log\left(\frac{p_1}{1-p_1}/\frac{p_2}{1-p_3}\right)$$

We need to know both p1 and  $\beta$  to interpret this.

#### The Meaning of a Logit Coefficient

$$Log~Odds~Ratio = Log\left(\frac{p_1}{1-p_1}/_{\frac{p_2}{1-p_2}}\right)$$
 If p1 = 0.5, Odds Ratio = 1

With  $\beta=0.01347$ , we multiply 1 by  $e^{0.01347}$ .

The new odds ratio is 1.013561, which means p=0.5033674



# Seed Predators



http://denimandtweed.com

```
A Quick Note on Within and Transformation
```

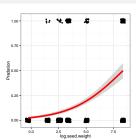
log.seed.weight

0.25

#### The $\mathsf{GLM}$

```
seeds <- within(seeds, {
  log.seed.weight <- log(seed.weight)
})</pre>
```

#### Fitted Seed Predation Plot



Residuals vs Fitted

Normal Q-Q

Diagnostics Look Odd Due to Binned Nature of the Data

#### Binned Residuals Should Look Spread Out

