

Poisson Regression



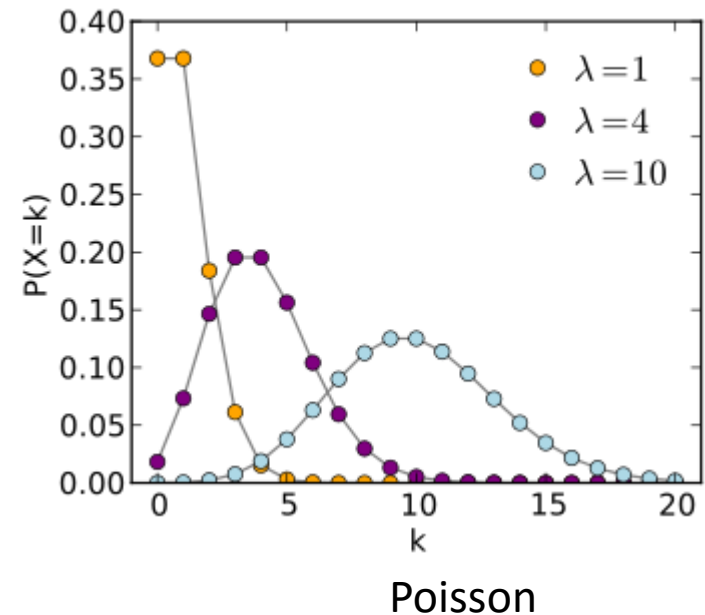
Some slides from Craig Burkett

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Elephants example

- 41 male elephants from Kenyan National Park
 - Followed for 8 years
- Interested in relationship between age & number of successful matings
- (Elephants, in R)
- Can't use linear regression because responses are not normal
 - Unless matings happen *really* often



Elephants Example - Binomial

- Can't use binomial response because the number of trials isn't fixed
 - What does a fixed number of trials entail?

Generalized Linear Models -- Review

$$g(\mu) = X\beta$$
$$Y \sim \text{dist}_\mu. Y^{(i)} \sim \text{dist}_\mu(g^{-1}(X^{(i)}\beta))$$

$g(\mu)$ is called a **Link Function**. The distribution of the Y 's is the family

$g(\mu) = \mu$ is linear regression – Identity Link

$g(\mu) = \log(\mu)$ is Log Link – the default for Poisson regression

$g(\mu) = \log\left(\frac{\mu}{1-\mu}\right)$ is Logistic regression – Logit Link

Estimating GLMs

- Find a β such that

$$\prod_i \text{dist}_\mu(g^{-1}(X^{(i)}\beta))$$

is maximized

Response

- Counts, so model with Poisson distribution

$$P(Y = y) = \frac{\mu^y e^{-\mu}}{y!}$$

$$E[Y] = \text{Var}[Y] = \mu$$

- Poisson regression model

$$\log(\mu) = \beta_0 + \beta_1 X_1 + \cdots + \beta_p X_p$$

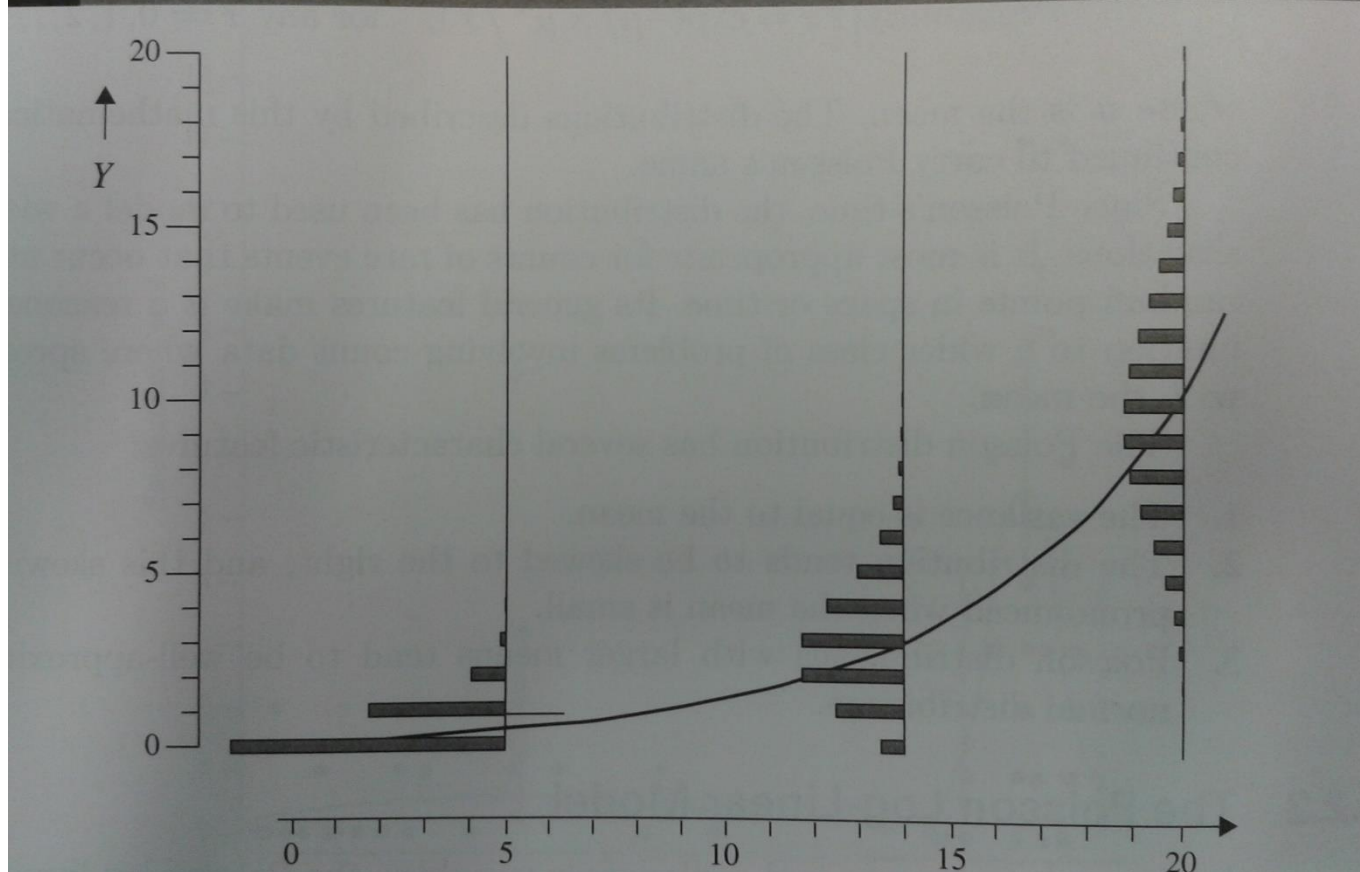
Interpretation

- $\mu = \exp(\beta_0 + \beta_1 X_1 + \dots + \beta_p X_p)$
- Increasing x_j by one unit with other predictors held constant, increases mean of response by factor e^{β_j}
- Inference (i.e, CI's for the β s)
 - Wald and LRT as with logistic

Visualizing the SD

2.5

A representation of a log-linear model in which the distribution of Y (as a function of X) is Poisson with mean μ and $\log(\mu) = -1.7 + 0.20X$; the histograms are the Poisson distribution at three values of X



Conclusions

- Does the average number of matings increase with age?
 - Yes
- For every additional year of age, the expected number of matings increases by
 - $\exp(0.0687)=1.07$