

## Poisson Reg. one Categ. predictor - crab data

$\mu$  = mean # of satellites per female crab

$$\log \hat{\mu} = 1.4069 - 0.2146 X_M - 0.6061 X_{DM} - 0.6918 X_D$$

$$X_C = \begin{cases} 1 & \text{color } C \\ 0 & \text{else} \end{cases}$$

exponentiated  $\rightarrow 4.083 \quad 0.807 \quad 0.845 \quad 0.801$

Int: The predicted number of satellites for light medium female crabs is 4.083 males per crab. (LM)

$X_{DM}$ :

$$\frac{\log(\hat{\mu})}{\text{LM}} = 1.4069$$

$$M \quad 1.4069 - 0.2146$$

$$DM \quad 1.4069 - 0.6061$$

$$D \quad 1.4069 - 0.6918$$

$$\hat{E}(Y|LM) = e^{1.4069}$$

$$\hat{E}(Y|DM) = e^{1.4069 - 0.6061}$$

$$= e^{1.4069} \cdot e^{-0.6061}$$

Interpret: The predicted average number of satellites for dark medium female crabs is 4.083 lower than that of light medium crabs.

Treat color as quantitative?

$$\log \hat{\mu} = 1.714 - 0.273 X \quad X = \text{color}$$

What % change in  $\hat{\mu}$  for DM compared to LM?

$$\frac{\hat{E}(Y|DM)}{\hat{E}(Y|LM)} = \frac{e^{1.714 - 0.273(3)}}{e^{1.714 - 0.273(1)}} = e^{-2(0.273)} \approx 0.579$$

$$= \begin{cases} 1 & LM \\ 2 & M \\ 3 & DM \\ 4 & D \end{cases}$$

## Poisson Regression with Rates

Response = count over some time frame / 1 crab /

- If time frames (ref. unit) changes across observations → can't compare counts directly,

Model rate parameter:  $\lambda = \frac{\mu}{t}$  ← mean count  
← unit time

Often clinical trials:  $t$  = "person-year"

Example: # of hip fractures at a nursing home

$\lambda$  = average number of hip fractures per person-year

$T$  = average # of residents at nursing home per year

$\lambda T$  = # of hip fractures at nursing home per year

$$\mu = E(Y) = \lambda T$$

Poisson Model:  $\underbrace{\mu}_{\sim} = \lambda T \Leftrightarrow \lambda = \frac{\mu}{T}$

$$\log(\lambda) = \log\left(\frac{\mu}{T}\right)$$

Want to model this with our linear predictor

$$= \beta_0 + \beta_1 X_1 + \dots + \beta_k X_k$$

$$\Leftrightarrow \log(\mu) = \beta_0 + \beta_1 X_1 + \dots + \beta_k X_k + \log T$$

$$\Leftrightarrow \log(\lambda) = \beta_0 + \beta_1 X_1 + \dots + \beta_k X_k \quad \text{offset term}$$