

## SMGT 432: Bradley-Terry Lecture

### Announcements:

1. Rice Soccer home opener vs. UH tomorrow night!
2. Today is a lecture (points opportunity!)
3. Assignment #1 is due next Friday

### Questions:

1. What distinguishes the Bradley-Terry model from other regression models?
2. What are one strength and one weakness of the Bradley-Terry model?
3. How can you avoid over fitting when estimating a Bradley-Terry model?

**Story** first time trying to estimate this model as a grad student

**Pause** Has anyone heard of the Bradley-Terry model?

### Bradley-Terry Model

$$Y \sim \text{Normal}(\eta, \sigma^2)$$

$$\eta = \alpha + \beta_{\text{HOME}} - \beta_{\text{AWAY}}$$

$$\vec{y} = \begin{bmatrix} 23 \\ -4 \\ \dots \\ 1 \\ 9 \end{bmatrix} \quad X = \begin{bmatrix} 1 & 1 & 0 & \dots & 0 & -1 \\ 1 & -1 & 0 & \dots & 1 & 0 \\ \dots & & & & & \\ 1 & 0 & 0 & \dots & 0 & 1 \\ 1 & 0 & 0 & \dots & 0 & 0 \end{bmatrix}$$

### Rasch Model

$$Y \sim \text{Normal}(\eta, \sigma^2)$$

$$\eta = \alpha + \beta_{\text{OFFENSE}} + \gamma_{\text{DEFENSE}}$$

$$\vec{y} = \begin{bmatrix} 96 \\ 87 \\ \dots \\ 101 \\ 91 \end{bmatrix} \quad X = \begin{bmatrix} 1 & 1 & 0 & \dots & 0 & 0 & 1 & \dots & 0 \\ 1 & 0 & 1 & \dots & 0 & 1 & 0 & \dots & 0 \\ \dots & & & & & & & & \\ 1 & 0 & 0 & \dots & 1 & 0 & 1 & \dots & 0 \\ 1 & 0 & 0 & \dots & 0 & 0 & 0 & \dots & 1 \end{bmatrix}$$

**Emphasize** What makes BT unique is that the design matrix  $X$  is sparse and always 0/1/-1.

**Pause** Confess love for BT model

How does it work?

### Maximum Likelihood

$$\mathcal{L}(\alpha, \beta) = \prod_{i=1}^n \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(y_i - \eta_i)^2}{2\sigma^2}}$$

$$\ell(\alpha, \beta) = \log(\mathcal{L}(\alpha, \beta)) = -\frac{n}{2} \log(2\pi\sigma^2) + \sum_{i=1}^n \frac{-(y_i - \eta_i)^2}{2\sigma^2}$$

$$\begin{aligned} \hat{\alpha}, \hat{\beta} &= \arg \max_{\alpha, \beta} \sum_{i=1}^n -(y_i - \eta_i)^2 \\ &= \arg \min_{\alpha, \beta} \sum_{i=1}^n \end{aligned}$$

$$\text{Ex 1: } \alpha = 3 \quad \beta_{\text{Rice}} = 4 \quad \beta_{\text{UH}} = 0$$

Rice beats UH by 7.

$$\ell(\alpha, \beta) = -(7 - (3+4-0))^2 = 0$$

Rice beats UH by 10.

$$\ell(\alpha, \beta) = -(10 - (3+4-0))^2 = -9$$

### Generalizations

Generalized Linear Models (GLMs):

$$Y \sim \text{Bernoulli} \left( \frac{e^\eta}{1+e^\eta} \right)$$

$$Y \sim \text{Poisson} (e^\eta)$$

Ex. 2: (H) AFC Richmond 3 - 2

$$\mathcal{L}(\eta) = P(Y=3) = \frac{\lambda^3 e^{-\lambda}}{3!} \quad \lambda = e^\eta$$

$$\ell(\eta) = \log\left(\frac{\lambda^3 e^{-\lambda}}{3!}\right) = 3 \log \lambda - \lambda - \log 3!$$

$$\frac{\partial}{\partial \lambda} \ell(\eta) = 3/\lambda - 1$$

$$\frac{\partial}{\partial \lambda} \ell(\eta) = 0 \Rightarrow \lambda = 3 \quad \Rightarrow \eta = \log 3 = 1.1$$

$$\eta = \alpha + \beta_{\text{Richmond}} + \gamma_{\text{WH}}$$

$$1.1 = 0.37 + \beta_{\text{Richmond}} + 0.35$$

$$\Rightarrow \beta_{\text{Richmond}} = 0.38$$

**Pause** This means we're estimating Richmond to be the strongest offense in the league. How do we feel about this?

## Preventing Overfitting

1. Random/Mixed-Effect Model (Traditional Statistics)
2. Bayesian Regression (Bayesian Statistics)
3. Regularization (Statistical Machine Learning)