

# **Week Eleven**

## **Last Week**

- Exam Recap
- Probability Distributions for Count Data
- Count Regression

## **This Week: GLMs for Count Data & Multicategory data**

- Bike Data

## **Next Week: Multicategory Regression**

---

## Multicategory Logit Models

Recall logistic regression for binary model

1. Random component / Probability distribution  $[0, 1]$

$$Y \sim \text{Binomial}(n, \pi) \text{ or } Y \sim \text{Bernoulli}(\pi)$$

2. Systematic component / Linear combination of predictors / functional form of predictors

$$\text{logit}(\pi) = XB \rightarrow B_0 + B_1 X_1 + B_2 X_1^2 + B_3 X_3$$

$$\log\left(\frac{\pi}{1-\pi}\right) = XB$$

3. Link function

1. Random component for multicategory data?

$$\vec{Y} = (Y_1, Y_2, \dots, Y_J) \quad \vec{\pi} = (\pi_1, \pi_2, \dots, \pi_J)$$

$$\vec{Y} \sim \text{Multinomial}(N, \vec{\pi}) \quad \sum_j \pi_j = 1 \quad \pi_j \in [0, 1]$$

2. Functional form  $\rightarrow \vec{X}\vec{B}$

2

3. Link function?

### Baseline-Category Logits

Let  $J$  be a baseline, then our baseline logit

is  $\log\left(\frac{\pi_j}{\pi_J}\right) \rightarrow$

$$J=2$$

$$\pi_1 + \pi_2 = 1$$

$$\log\left(\frac{\pi_j}{\pi_J}\right) = \alpha_j + \beta_j x$$

$$\pi_2 = 1 - \pi_1$$

$$\log\left(\frac{\pi_1}{\pi_2}\right) = \log\left(\frac{\pi_1}{1-\pi_1}\right)$$

$$\log\left(\frac{\pi_1}{\pi_3}\right) = \alpha_1 + \beta_1 x$$

require  $J-1$  equations

AND

$$\log\left(\frac{\pi_2}{\pi_3}\right) = \alpha_2 + \beta_2 x$$

$$\log\left(\frac{\pi_1}{\pi_2}\right) = \log\left(\frac{\pi_1 / \pi_J}{\pi_2 / \pi_J}\right) = \log\left(\frac{\pi_1}{\pi_J}\right) - \log\left(\frac{\pi_2}{\pi_J}\right)$$

$$= (\alpha_1 + \beta_1 x) - (\alpha_2 + \beta_2 x) \Rightarrow (\alpha_1 - \alpha_2) +$$

$\Rightarrow$  odds ratio

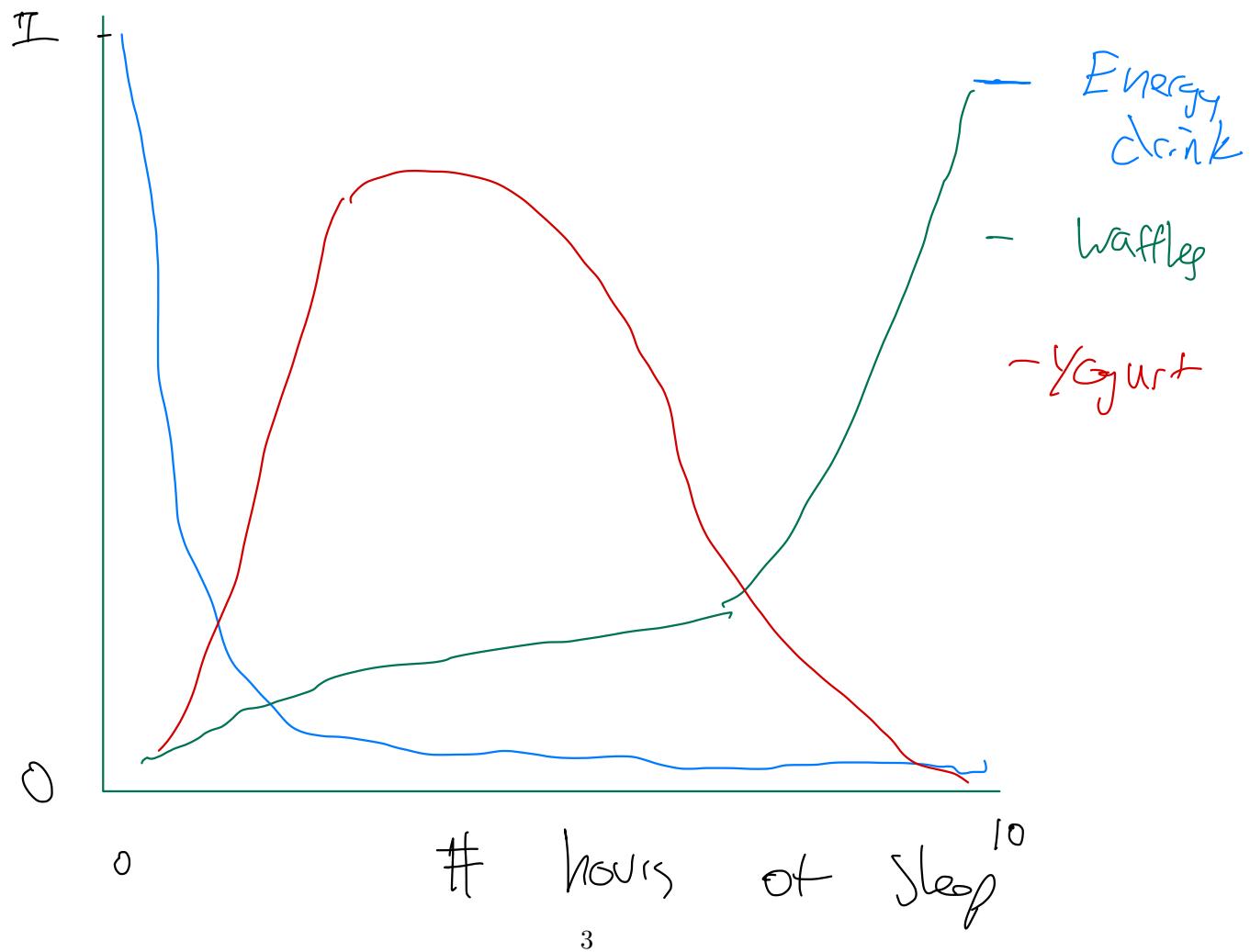
3

$$(\beta_1 - \beta_2) x$$

### Baseline-Category Logits

We want estimate  $\hat{\pi}(x)$

$$\hat{\pi}_i(x) = \frac{\exp(\alpha_i + B_i x)}{\sum_{j=1}^J \exp(\alpha_j + B_j x)} \rightarrow \alpha_J + B_J = 0$$



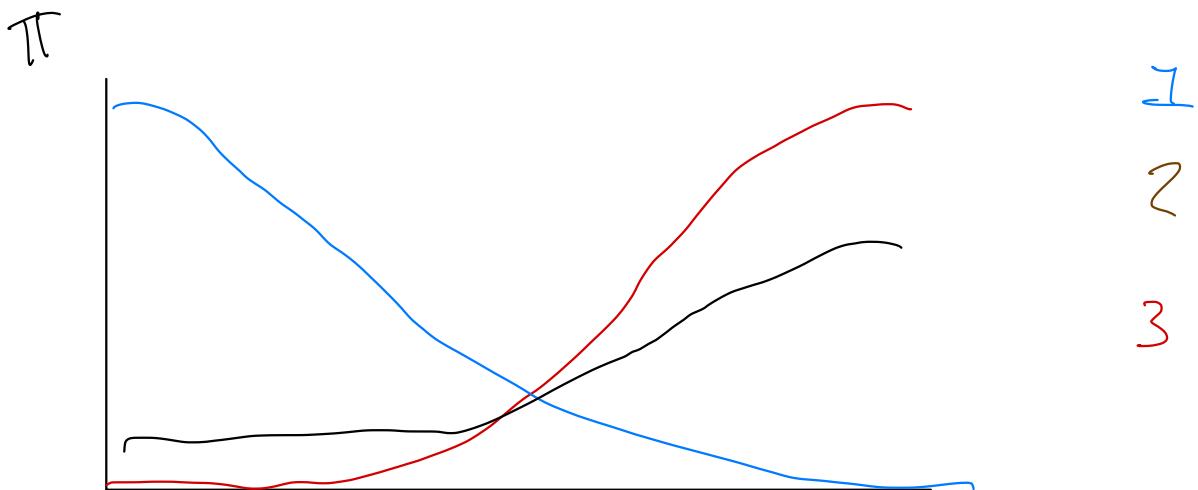
Ordinal Data

$$\Pr[Y \leq j] = \pi_1 + \pi_2 + \dots + \pi_j$$

Cumulative probabilities

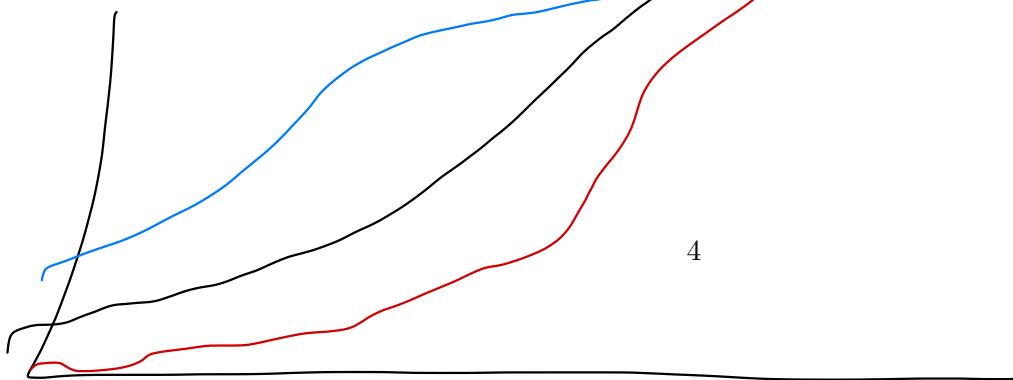
### 1. Multinomial

$$\text{logit}(\Pr[Y \leq j]) = \alpha_j + \beta X$$



grams of  
butter

3-2-1



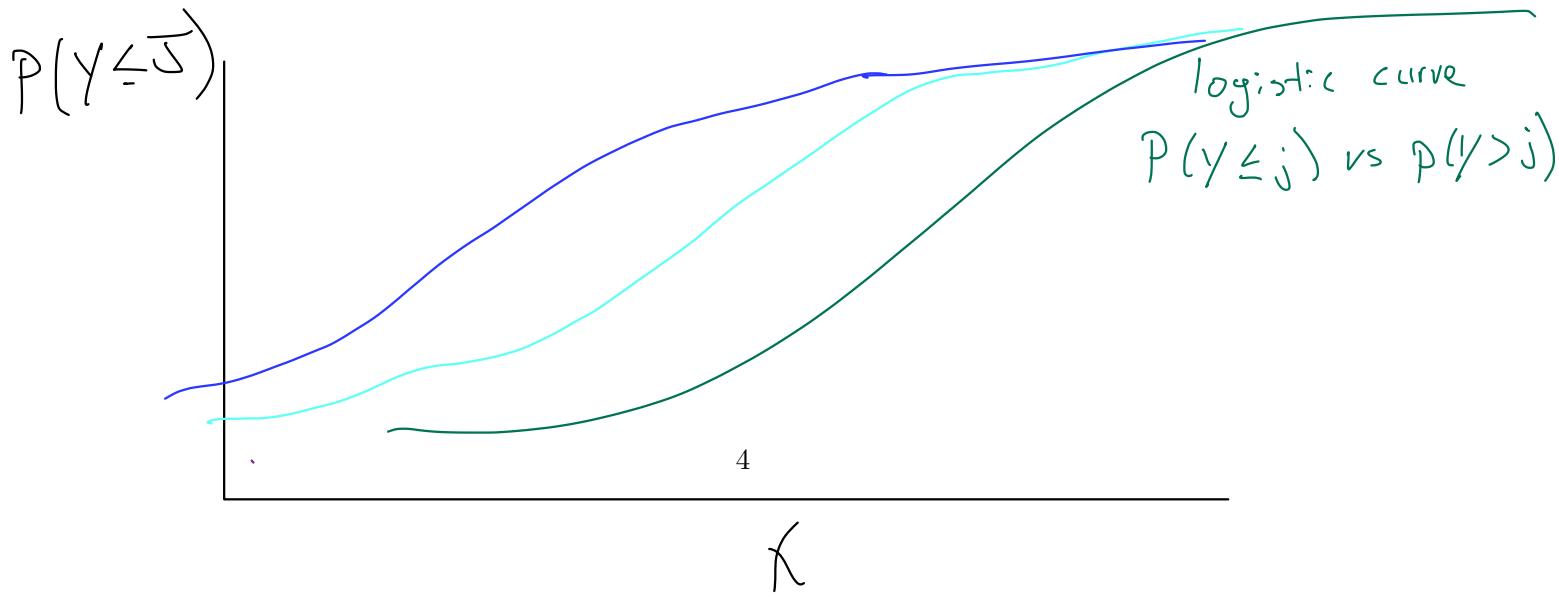
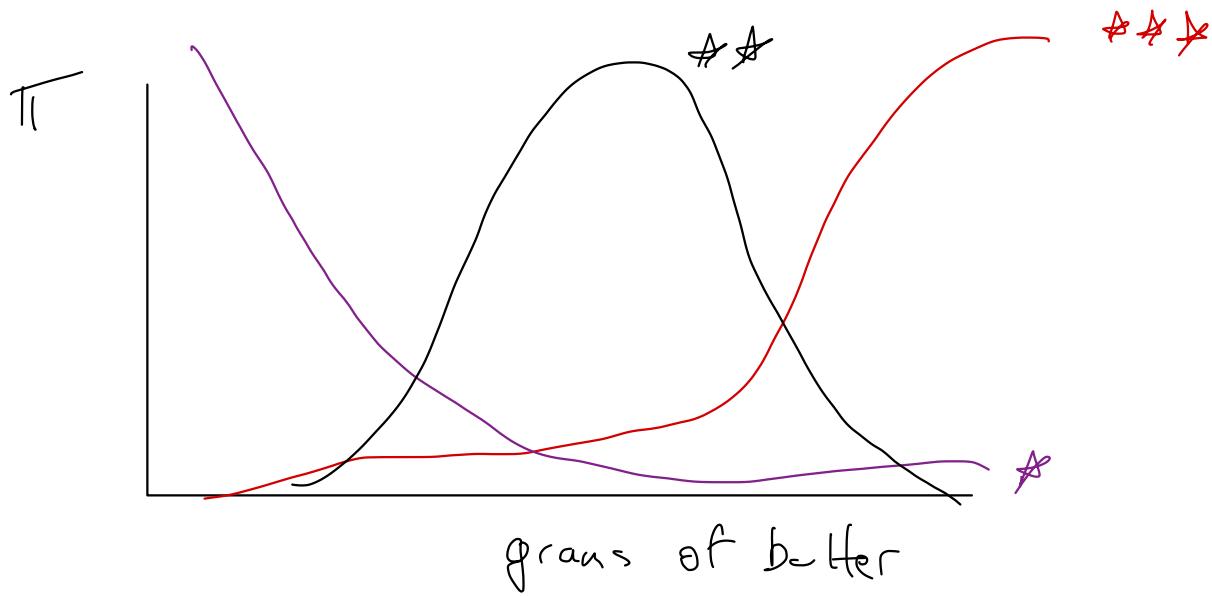
4

Ordinal Data

5

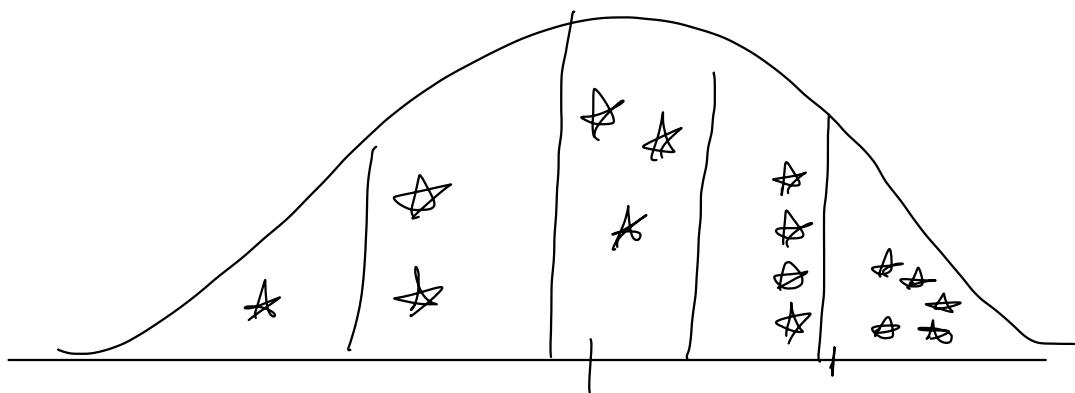
$$P_c [Y \leq j] = \pi_1 + \pi_2 + \dots + \pi_j$$

Cumulative probabilities

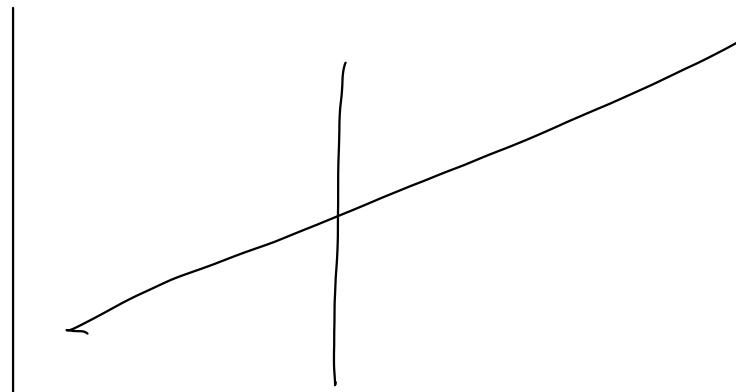


### Latent variable interpretation

- ↳ unmeasured (hidden) continuous variable
- ↳ Normal continuous



Probit  $\rightarrow$  CDF of Normal



better