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STAT 2132
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Homework 5

$$P(Y_i = j) = \pi_{ij} > 0, \quad j = 1, 2, 3, \quad i = 1, \dots, n$$

$$\log \left(\frac{\pi_{i1}}{\pi_{i2} + \pi_{i3}} \right) = \alpha_1 - \beta x_i \quad (1)$$

$$\log \left(\frac{\pi_{i1} + \pi_{i2}}{\pi_{i3}} \right) = \alpha_2 - \beta x_i \quad (2)$$

(a) From (1)

$$\Rightarrow \frac{\pi_{i1}}{\pi_{i2} + \pi_{i3}} = e^{\alpha_1 - \beta x_i}$$

$$\Rightarrow \pi_{i1} = (\pi_{i2} + \pi_{i3}) e^{\alpha_1 - \beta x_i}$$

Substitute this into (2)

$$\Rightarrow \frac{(\pi_{i2} + \pi_{i3}) e^{\alpha_1 - \beta x_i} + \pi_{i2}}{\pi_{i3}} = e^{\alpha_2 - \beta x_i}$$

$$\Rightarrow \pi_{i2} (e^{\alpha_1 - \beta x_i} + 1) + \pi_{i3} e^{\alpha_1 - \beta x_i} = \pi_{i3} e^{\alpha_2 - \beta x_i}$$

$$\Rightarrow \pi_{i2} = \frac{\pi_{i3} (e^{\alpha_2 - \beta x_i} - e^{\alpha_1 - \beta x_i})}{e^{\alpha_1 - \beta x_i} + 1}$$

$$= \frac{\pi_{i3} e^{-\beta x_i} (e^{\alpha_2} - e^{\alpha_1})}{e^{\alpha_1 - \beta x_i} + 1}$$

$$= \frac{\pi_{i3} (e^{\alpha_2} - e^{\alpha_1})}{e^{\alpha_1} + e^{\beta x_i}}$$

From (2) and $\pi_{i3} = 1 - (\pi_{i1} + \pi_{i2})$ we also have

$$\log \left(\frac{1 - \pi_{i3}}{\pi_{i3}} \right) = \alpha_2 - \beta x_i$$

$$\Rightarrow \frac{1 - \pi_{i3}}{\pi_{i3}} = e^{\alpha_2 - \beta x_i}$$

$$\Rightarrow \frac{1}{\pi_{i3}} = e^{\alpha_2 - \beta x_i} + 1$$

$$\Rightarrow \pi_{i3} = \frac{1}{e^{\alpha_2 - \beta x_i} + 1}$$

Substitute this into (3)

$$\Rightarrow \pi_{i2} = \frac{e^{\alpha_2} - e^{\alpha_1}}{(e^{\alpha_1} + e^{\beta x_i})(e^{\alpha_2 - \beta x_i} + 1)}$$

$$(b) \quad \pi_{i2} > 0, \quad (e^{\alpha_1} + e^{\beta x_i})(e^{\alpha_2 - \beta x_i} + 1) > 0$$

$$\Rightarrow e^{\alpha_2} - e^{\alpha_1} > 0$$

$$\Rightarrow e^{\alpha_2} > e^{\alpha_1} \Rightarrow \alpha_2 > \alpha_1$$

When $\pi_{i2} = 0$, that means y_i takes on only values 1 and 3

$$\Rightarrow e^{\alpha_2} - e^{\alpha_1} = 0 \Rightarrow \alpha_2 = \alpha_1$$

Therefore, $\alpha_2 \geq \alpha_1$

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2/22/2021

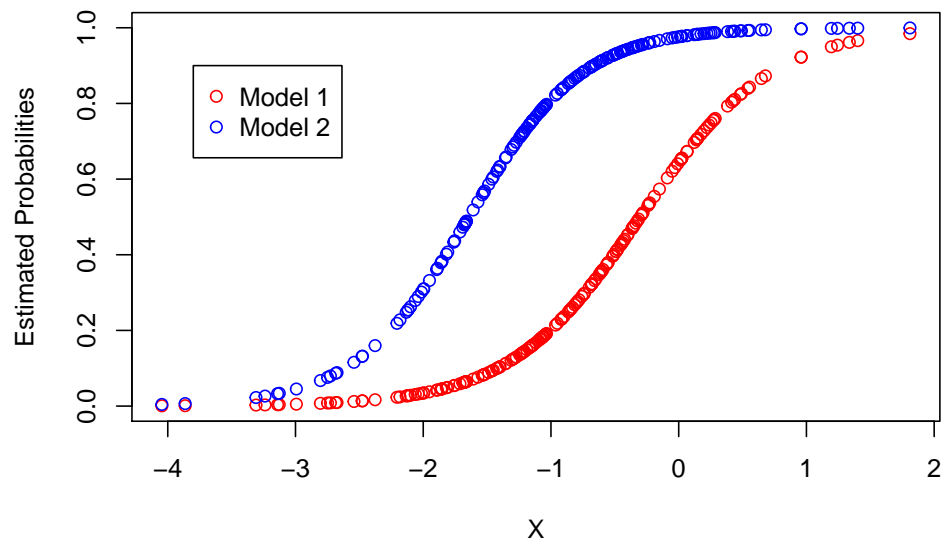
(c)

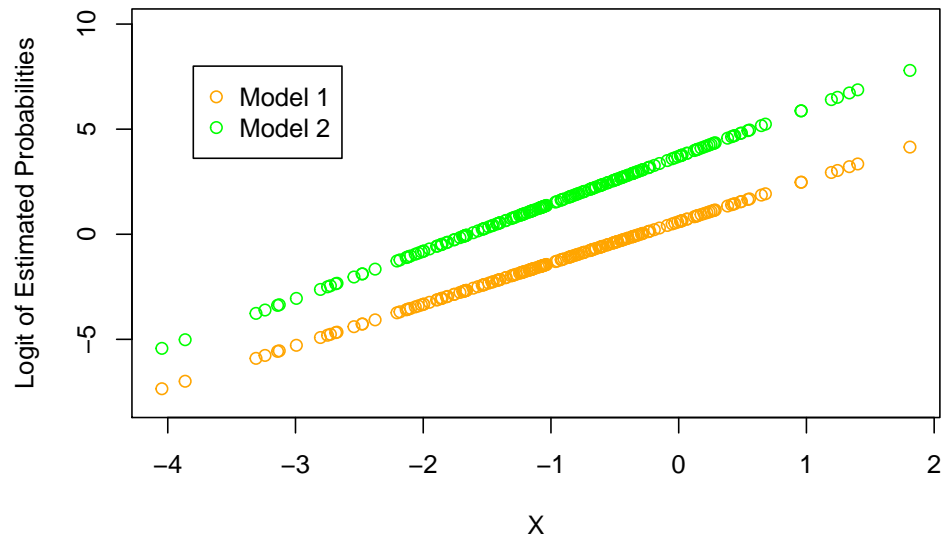
The fitted models are

(Mi) $\text{logit}(\hat{\pi}_{i1}) = 0.5932 + 1.9634x_i$

(Mii) $\text{logit}(\hat{\pi}_{i1} + \hat{\pi}_{i2}) = 3.707 + 2.258x_i$

(i) & (ii) Below is the plot of estimated probabilities from (Mi) and the estimated probabilities from (Mii) as a function of x_i (red and blue), as well as the logit of them as a function of x_i (orange and green).





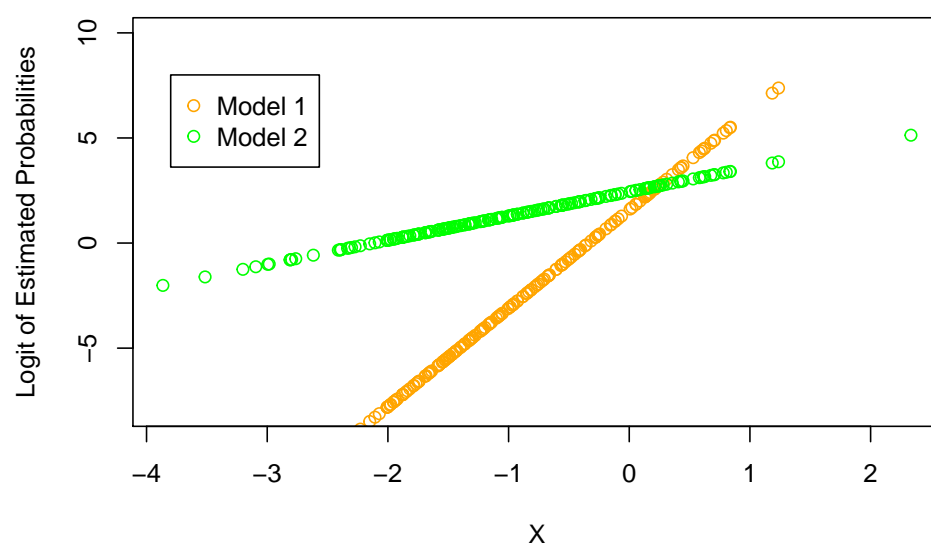
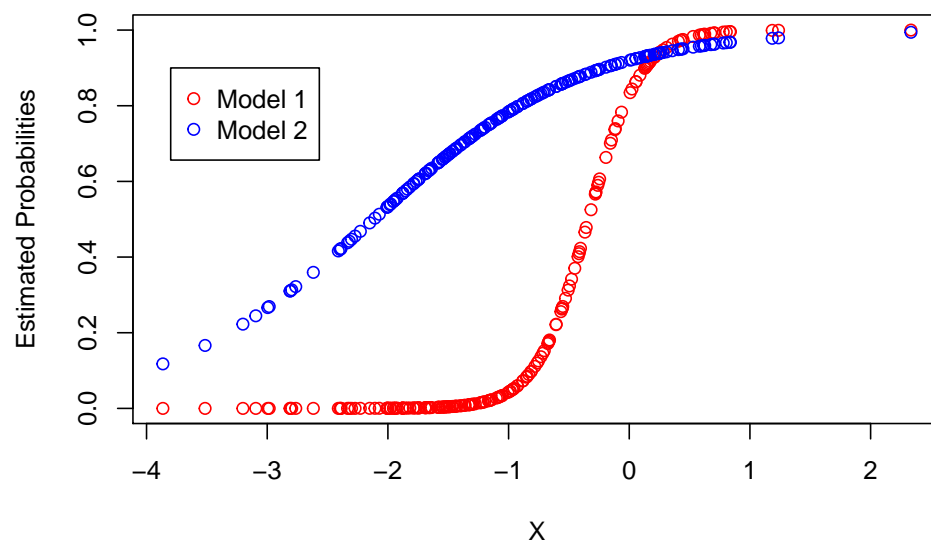
(d)

Using another sample, the fitted models are

(Mi) $\text{logit}(\hat{\pi}_{i1}) = 1.584 + 4.684x_i$

(Mii) $\text{logit}(\hat{\pi}_{i1} + \hat{\pi}_{i2}) = 2.441 + 1.153x_i$

(i) & (ii) Below is the plot of estimated probabilities from (Mi) and the estimated probabilities from (Mii) as a function of x_i (red and blue), as well as the logit of them as a function of x_i (orange and green).



(e)

From the plots in parts (c) and (d), I am more comfortable fitting the proportional odds model to Q1d.txt sample because in Q1d.txt data, the plots of two different models cross each other. In the proportional odds model given, we see that the two models we want to fit have the same slope β . Therefore we would expect the plots of two models to be kind of parallel in Q1c.txt, not crossing like Q1d.txt

proportional odds will be more appropriate for Q1c bc of the plots