

$$1. y_0 = \arg \max_y P(y_0=y|\pi) \prod_{d=1}^D P_d(x_{0,d}|\theta_y^{(d)})$$

$$\Rightarrow \hat{\pi}, \hat{\theta}_y^{(1)}, \hat{\theta}_y^{(2)} = \arg \max_{\pi, \theta_y^{(1)}, \theta_y^{(2)}} \sum_{i=1}^n \ln P(y_i|\pi) + \sum_{i=1}^n \ln P(x_{i,1}|\theta_y^{(1)}) + \sum_{i=1}^n \ln P(x_{i,2}|\theta_y^{(2)})$$

$$\text{and } \because P(y|\pi) \sim \text{Bernoulli}(y|\pi)$$

$$\therefore P(y|\pi) = \pi^{y_i} (1-\pi)^{(1-y_i)}$$

$$\therefore P_1 \sim \text{Bernoulli}(x_{0,1}|\theta_y^{(1)})$$

$$\therefore P(x_{i,1}|\theta_y^{(1)}) = (\theta_y^{(1)})^{x_{0,1}} (1-\theta_y^{(1)})^{1-x_{0,1}}$$

$$\therefore P_2 \sim \text{Pareto}(x_{0,2}|\theta_y^{(2)})$$

$$\therefore P(x_{i,2}|\theta_y^{(2)}) = \theta_y^{(2)} (x_{0,2})^{-(\theta_y^{(2)}+1)}$$

$$\text{Let } L = \sum_{i=1}^n [y_i \ln \pi + (1-y_i) \ln(1-\pi)] + \sum_{i=1}^n [x_{i,1} \ln \theta_y^{(1)} + (1-x_{i,1}) \ln(1-\theta_y^{(1)})] + \sum_{i=1}^n [-(\theta_y^{(2)}+1) \ln x_{i,2} + \ln \theta_y^{(2)}]$$

$$(a) \nabla_{\pi} L = \sum_{i=1}^n [\frac{y_i}{\pi} - \frac{1-y_i}{1-\pi}]$$

$$\text{let } \nabla_{\pi} L = 0, \text{ then } \sum_{i=1}^n [\frac{y_i}{\pi} - \frac{1-y_i}{1-\pi}] = 0$$

$$\Rightarrow \sum_{i=1}^n [(1-\pi)y_i - \pi(1-y_i)] = 0 \Rightarrow \pi = \frac{1}{n} \sum_{i=1}^n y_i$$

$$\therefore \hat{\pi} = \frac{1}{n} \sum_{i=1}^n y_i$$

$$(b) \nabla_{\theta_y^{(1)}} L = \sum_{i=1}^n [\frac{x_{i,1}}{\theta_y^{(1)}} + \frac{1-x_{i,1}}{1-\theta_y^{(1)}} \times (-1)] \mathbb{1}(y_i=y)$$

$$\text{let } \nabla_{\theta_y^{(1)}} L = 0, \text{ then } \sum_{i=1}^n [\frac{x_{i,1}}{\theta_y^{(1)}} - \frac{1-x_{i,1}}{1-\theta_y^{(1)}}] \mathbb{1}(y_i=y) = 0$$

$$\Rightarrow \sum_{i=1}^n [(1-\theta_y^{(1)})x_{i,1} - (1-x_{i,1})\theta_y^{(1)}] \mathbb{1}(y_i=y) = 0 \Rightarrow \theta_y^{(1)} = \frac{1}{n_y} \sum_{i=1}^n \mathbb{1}(y_i=y) x_{i,1}$$

$$\therefore \hat{\theta}_y^{(1)} = \frac{1}{n_y} \sum_{i=1}^n \mathbb{1}(y_i=y) x_{i,1} \quad \text{where } n_y \text{ is the number of } y_i=y \text{ (} y=0,1 \text{)}$$

$$(c) \nabla_{\theta_y^{(2)}} L = \sum_{i=1}^n [\mathbb{1}(y_i=y) \cdot (\frac{1}{\theta_y^{(2)}} - \ln x_{i,2})]$$

$$\text{Let } \nabla_{\theta_y^{(2)}} L = 0 \text{ then } \frac{1}{\theta_y^{(2)}} = \frac{n_y}{\sum_{i=1}^n \mathbb{1}(y_i=y) \ln(x_{i,2})}$$

$$\therefore \hat{\theta}_y^{(2)} = \frac{n_y}{\sum_{i=1}^n \mathbb{1}(y_i=y) \ln(x_{i,2})}$$

2.

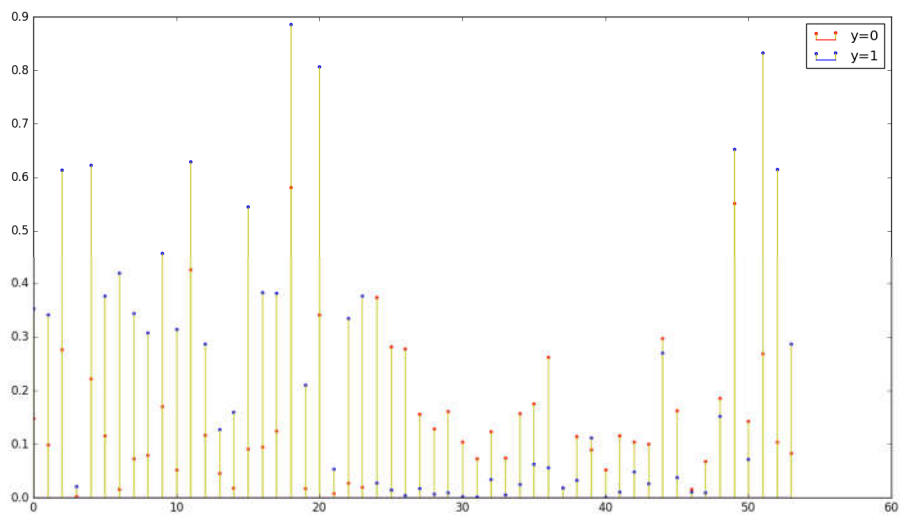
(a)

+-----+-----+-----+			
	/	actual y=0	actual y=1

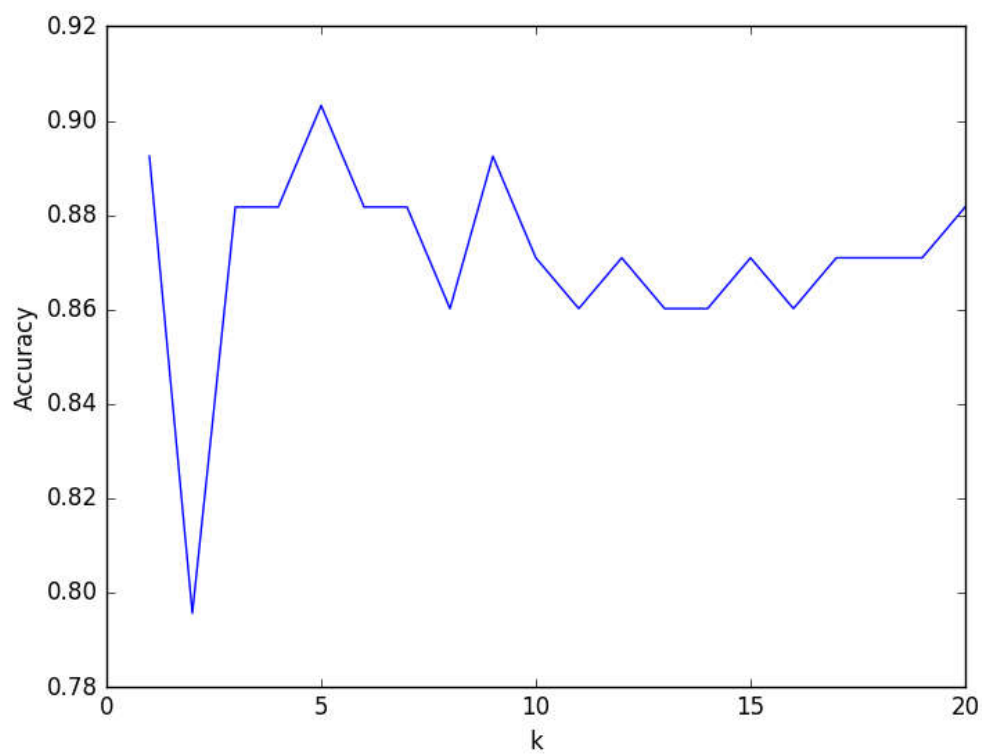
predicted y=0	54	6
predicted y=1	2	31

0.913978494624

(b)



(c)



(d)

