COMS 4771 Machine Learning (Spring 2015) Problem Set #1

Anji Zhao - az2324@columbia.edu Discussants: mo2454

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1. Given
$$S = \left\{ x_{i,1:d}, k_i \right\}_{i=1}^n$$

likelihood
$$L = \prod_{i=1}^{n} \prod_{j=1}^{d} \prod_{k \in \mathcal{Y}} p_{j,k}^{x_{ij}} (1 - p_{j,k})^{(1-x_{ij})}$$

$$\ln(L) = \sum_{i=1}^{n} \sum_{j=1}^{d} \sum_{k \in \mathcal{Y}} (x_{ij} \ln(p_{jk}) + (1 - x_{ij}) \ln(1 - p_{jk}))$$

$$\ln(L) = \sum_{i=1}^{n} \sum_{j=1}^{d} \mathbb{I}\{k_i = k\} (\mathbb{I}\{x_j = 1\} \ln(p_{jk}) + (1 - \mathbb{I}\{x_j = 1\}) + \mathbb{I}\{x_j = 1\})$$

$$\frac{\partial \ln(L)}{\partial p_{j,k}} = \sum_{i=1}^{n} \sum_{j=1}^{d} \mathbb{I}\{k_i = k\} \left(\frac{\mathbb{I}\{x_j = 1\}}{p_{jk}} - \frac{1 - \mathbb{I}\{x_j = 1\}}{1 - p_{jk}}\right)$$

$$0 = \sum_{i=1}^{n} \sum_{j=1}^{d} \mathbb{I}\{k_i = k\} \left(\frac{\mathbb{I}\{x_j = 1\}}{p_{jk}} - \frac{\mathbb{I}\{x_j = 0\}}{1 - p_{jk}}\right)$$

$$\hat{p_{jk}} \sum_{i=1}^{n} \sum_{j=1}^{d} \mathbb{I}\{k_i = k\} \mathbb{I}\{x_j = 0\} = (1 - p_{jk}) \sum_{i=1}^{n} \sum_{j=1}^{d} \mathbb{I}\{k_i = k\} \mathbb{I}\{x_j = 1\}$$

$$\hat{p_{jk}} \left(\sum_{i=1}^{n} \sum_{j=1}^{d} \mathbb{I}\{k_i = k\} \mathbb{I}\{x_j = 1\}\right)$$

$$\hat{p_{jk}} = \frac{\sum_{i=1}^{n} \sum_{j=1}^{d} \mathbb{I}\{k_i = k\} \mathbb{I}\{x_j = 1\}}{\sum_{i=1}^{n} \sum_{j=1}^{d} \mathbb{I}\{k_i = k\}}$$

2. If there are only two newsgroups, we can map $\{0,1\} \mapsto \{1,2\} \mapsto \{-1,+1\}$.

$$f(x) = 1 \text{ when } \Pr(Y = 1 \mid X = x) > \Pr(Y = 2 \mid X = x).$$

$$\pi_1 \prod_{j=1}^d p_{j1}^{x_j} (1 - p_{j1})^{(1-x_j)} > \pi_2 \prod_{j=1}^d p_{j2}^{x_j} (1 - p_{j2})^{(1-x_j)}$$

$$\frac{\prod_{j=1}^d p_{j1}^{x_j} (1 - p_{j1})^{(1-x_j)}}{\prod_{j=1}^d p_{j2}^{x_j} (1 - p_{j2})^{(1-x_j)}} > \frac{\pi_2}{\pi_2}$$

$$\prod_{j=1}^d \left(\frac{p_{j1}}{p_{j2}}\right)^{x_j} \left(\frac{1 - p_{j1}}{1 - p_{j2}}\right)^{(1-x_j)} > \frac{\pi_2}{\pi_2}$$

$$\ln\left(\prod_{j=1}^d \left(\frac{p_{j1}}{p_{j2}}\right)^{x_j} \left(\frac{1 - p_{j1}}{1 - p_{j2}}\right)^{(1-x_j)}\right) > \ln\left(\frac{\pi_2}{\pi_2}\right)$$

$$\sum_{j=1}^d x_j \ln\left(\frac{p_{j1}}{p_{j2}}\right) + (1 - x_j) \ln\left(\frac{1 - p_{j1}}{1 - p_{j2}}\right) > \ln\frac{\pi_2}{\pi_2}$$

$$\left(\sum_{j=1}^d x_j \ln\left(\frac{p_{j1}}{p_{j2}}\right) + (1 - x_j) \ln\left(\frac{1 - p_{j1}}{1 - p_{j2}}\right)\right) - \left(\ln\frac{\pi_2}{\pi_2}\right) > 0$$

$$\sum_{j=1}^d x_j \left(\ln\left(\frac{p_{j1}}{p_{j2}}\right) - \ln\left(\frac{1 - p_{j1}}{1 - p_{j2}}\right)\right) + \ln\left(\frac{1 - p_{j1}}{1 - p_{j2}}\right) - \ln\frac{\pi_2}{\pi_2} > 0$$

$$\text{Let } w = \left(\ln\left(\frac{p_{j1}}{p_{j2}}\right) - \ln\left(\frac{1 - p_{j1}}{1 - p_{j2}}\right)\right) \text{ and } \theta > \ln\frac{\pi_2}{\pi_2} - \ln\left(\frac{1 - p_{j1}}{1 - p_{j2}}\right).$$

$$\text{Let } 2 \mapsto -1 \text{ and } 1 \mapsto +1. \text{ Then we have } f(x) = 1 \text{ when } (\langle w, x \rangle - \theta) > 0.$$

$$\text{or } f(x) = \text{sign}(\langle w, x \rangle - \theta)$$

3. training error = 0.22163test error = 0.3760