

HW4(第一题)

葛宇泽

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1 Q1

$$\begin{aligned} l(\beta) &= \sum_{i=1}^N \{y_i \log p(x_i; \beta) + (1 - y_i) \log (1 - p(x_i; \beta))\} \\ &= \sum_{i=1}^N \left\{ y_i \log \frac{p(x_i; \beta)}{1 - p(x_i; \beta)} + \log (1 - p(x_i; \beta)) \right\} \end{aligned}$$

$$\text{而: } p(x_i; \beta) = P(Y = 1 | X = x_i) = \frac{e^{x_i' \beta}}{1 + e^{x_i' \beta}}$$

$$1 - p(x_i; \beta) = \frac{1}{1 + e^{x_i' \beta}}$$

$$\begin{aligned} \text{代入得: } l(\beta) &= \sum_{i=1}^N \left\{ y_i \log e^{x_i' \beta} - \log (1 + e^{x_i' \beta}) \right\} \\ &= \sum_{i=1}^N \left\{ y_i x_i^\top \beta - \log (1 + e^{x_i' \beta}) \right\} \end{aligned}$$

2 Q2

$$\begin{aligned} \ell(\beta) &= \sum_{i=1}^N [y_i x_i^\top \beta - \log (1 + \exp (x_i^\top \beta))] \\ \frac{\partial \ell}{\partial \beta} &= \sum_{i=1}^n \left[y_i - \frac{\exp(x_i^\top \beta)}{1 + \exp(x_i^\top \beta)} \right] x_i = \sum_{i=1}^N [y_i - p(x_i; \beta)] x_i \end{aligned}$$

In the formula:

$$\begin{aligned} 1 - p(x_i; \beta) &= \frac{1}{1 + \exp(x_i^\top \beta)} \\ -\frac{\partial p(x_i; \beta)}{\partial \beta_j} &= -\frac{x_{ij} \exp(x_i^\top \beta)}{1 + \exp(x_i^\top \beta)} = -x_{ij} p(x_i; \beta) (1 - p(x_i; \beta)) \\ \frac{\partial p(x_i; \beta)}{\partial \beta} &= x_i p(x_i; \beta) (1 - p(x_i; \beta)) \end{aligned}$$

So:

$$\begin{aligned}\frac{\partial^2 l}{\partial \beta^\top \partial \beta}(k) &= - \sum_{i=1}^N \frac{\partial p(x_i; \beta)}{\partial \beta^\top} x_{ik} \\ &= - \sum_{i=1}^n x_i^\top p(x_i; \beta) (1 - p(x_i; \beta)) x_{ik}\end{aligned}$$

Which is the k-th row of $\frac{\partial^2 l}{\partial \beta^\top \partial \beta}$

$$\frac{\partial^2 l}{\partial \beta^\top \partial \beta} = - \sum_i p(x_i; \beta) (1 - p(x_i; \beta)) \begin{bmatrix} x_i^\top x_{i1} \\ \vdots \\ x_i^\top x_{im} \end{bmatrix} = - \sum_i p(x_i; \beta) (1 - p(x_i; \beta)) x_i x_i^\top$$

m is the dimension of x_i

代码部分见 HW4_NR.html

3 Q3

$$\begin{aligned}I(f(x^+) < f(x^-)) &= 1 - I(f(x^+) > f(x^-)) \\ l_{\text{rank}} &= \frac{1}{m^+ m^-} \sum_{x^+ \in D^+} \sum_{x^- \in D^-} (I(f(x^+) < f(x^-))) \\ &= \frac{1}{m^+ m^-} \sum_{x^+ \in D^+} \sum_{x^- \in D^-} (1 - I(f(x^+) > f(x^-))) \\ &= 1 - \frac{1}{m^+ m^-} \sum_{x^+ \in D^+} \sum_{x^- \in D^-} I(f(x^+) > f(x^-)) \\ 1 - l_{\text{rank}} &= \frac{1}{m^+ m^-} \sum_{x^+ \in D^+} \sum_{x^- \in D^-} I(f(x^+) > f(x^-)) \\ &= \sum_{x^- \in D^-} \frac{1}{m^-} \times \frac{1}{m^+} \sum_{x^+ \in D^+} I(f(x^+) > f(x^-))\end{aligned}$$

如图所示，我们将 ROC 曲线垂直于横轴分割为多个小长方形。在绘图过程中，如果选取的当前样本点为正例，则 ROC 曲线向上绘制 $1/m^+$ 而不向右绘制，故此类点与小长方形的个数无关。我们（由预测值从大到小）遍历每个负例，因为每个负例都会产生一个长为 $1/m^-$ 的小长方形。它的高为预测值比当前点更高的正例的个数 $\times 1/m^+$ 。（因为我们从预测值最高的点开始绘制，每遇到一个正例，就向上绘制制 $1/m^+$ ，直到当前点）将所有小长方形的面积相加，即为上式。

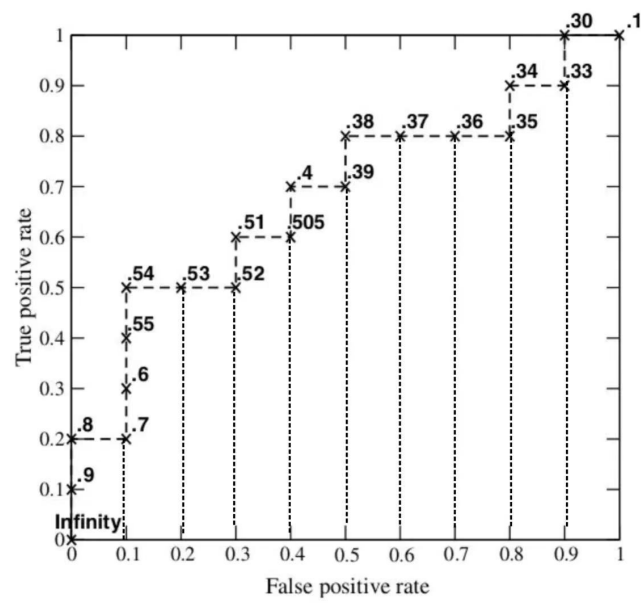


图 1: 将 ROC 曲线垂直于横轴分割为多个小长方体