

Lecture 1

$$\frac{P(y=1 | X=x)}{P(y=0 | X=x)} = \frac{\frac{P(X|y=1)P(y=1)}{P(X)}}{\frac{P(X|y=0)P(y=0)}{P(X)}}$$

posterior odds

$$\sim = LR \times \text{prior odds}$$

Bayes risk

$$r(\hat{y}^*(\cdot)) = \mathbb{E}_{x,y} [\ell(y, \hat{y}^*(x))]$$

$$= \mathbb{E}_x [\mathbb{E}_{y|x} [\ell(y, \hat{y}^*(x))]]$$

$$= \mathbb{E}_x [\min_a \mathbb{E}_{y|x} [\ell(y, a)]]$$

\Downarrow

$$r(\hat{y}(\cdot)) \geq r(\hat{y}^*(\cdot))$$



$$y \in \mathbb{R}, \quad \ell(y, \hat{y}) = (y - \hat{y})^2$$

$$\operatorname{argmin}_a \mathbb{E}_{y|x} [(y - a)^2] \equiv \hat{y}^*(x)$$

$$\frac{\partial}{\partial a} \mathbb{E}_{y|x} [y^2 - 2ya + a^2] = 0$$

$$2a - 2\mathbb{E}_{y|x} [y] = 0$$

$$a = \mathbb{E}_{y|x} [y]$$

Posterior mean



$$P(Y=1 \mid x_1, \dots, x_p) =$$

$$\frac{P(x_1, \dots, x_p \mid Y=1) P(Y=1)}{P(x_1, \dots, x_p)}$$

Naive Bayes assumption:

$$= \frac{\left[\prod_i P(x_i \mid Y=1) \right] P(Y=1)}{P(x_1, \dots, x_p)}$$

NOTE: The assumption does not imply
 $P(x_1, \dots, x_p) = \prod_i P(x_i)$

$$= \frac{\left[\prod_i P(x_i \mid Y=1) \right] P(Y=1)}{\sum_y \left[\prod_i P(x_i \mid Y=y) \right] P(Y=y)}$$

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