

The "True" Bayesian Approach.

MLE  $\Leftarrow$  maximum likelihood  $\hat{\theta}_{MLE} = \arg \max_{\theta} P(D; \theta)$

MAP  $\Leftarrow$  max a posteriori  $\hat{\theta}_{MAP} = \arg \max_{\theta} P(\theta|D)$

---

no single  $\theta$  captures our belief.

just use  $P(\theta|D)$

$$\hat{P}(X, Y) = \int_{\theta} P(X, Y|\theta) \cdot P(\theta|D) d\theta$$

$$= \int_{\theta} P(X, Y|\theta) \cdot \frac{P(D|\theta) \cdot P(\theta)}{P(D)} d\theta$$

## Naive Bayes

$$D = \{(x_1, y_1), (x_2, y_2), \dots, (x_n, y_n)\} \sim \mathcal{P}$$

$$P(D) = P((x_1, y_1), (x_2, y_2), \dots) = \prod_{i=1}^n P((x_i, y_i))$$

MLE estimate of  $\mathcal{P}$  is  $\hat{P}(x, y) = \frac{|\{i \mid x_i = x, y_i = y\}|}{n}$

$$\hat{P}(x, y) = \frac{1}{n} \sum_{i=1}^n I(x_i = x \text{ and } y_i = y)$$

$$I(\text{true}) = 1, \quad I(\text{false}) = 0.$$

Optimal classifier:  $h(x) = \arg \max_{y \in \mathcal{Y}} \hat{P}(x, y)$

$$h(x) = \arg \max_{y \in \mathcal{Y}} \frac{1}{n} \sum_{i=1}^n I(x_i = x \text{ and } y_i = y)$$

Suppose  $X = \{0, 1\}^d$ ,  $Y = \{-1, 1\}$ .

What is the size of the support of  $\hat{P}$ ?

That is, what is  $|X \times Y|$ ?  $2^{d+1}$

intractable

## Naive Bayes Assumption

If I let  $x_i^\alpha$  denote the  $\alpha$ th feature of the  $i$ th example, then

$x_i^\alpha$  and  $x_i^\beta$  are independent given  $y_i$ .

$x_i^1, x_i^2, \dots, x_i^d$  are all independent given  $y_i$ .

$$P(x_i, y_i) = P(x_i | y_i) P(y_i) = \left( \prod_{\alpha=1}^d P(x_i^\alpha | y_i) \right) P(y_i)$$

$$= \left( \prod_{\alpha=1}^d P(x_i^\alpha | y_i) \right) P(y_i)$$

$$P(D) = \prod_{i=1}^n \left( \prod_{\alpha=1}^d P(x_i^\alpha | y_i) \right) P(y_i)$$

---

Examples of tasks where Naive Bayes  $\wedge$  is reasonable?  
assumption

- movie recommender system
- text classification (genre/topic)
- demographic info  $\Leftarrow$  search
- dating app swipes  $\Rightarrow$  secret personal data

# Categorical Naive Bayes

$$x \in \{0, 1\}^d$$

~~$$P(x^\alpha | y) = \theta$$~~

$$P(x_i^\alpha | y=c) = \theta_{\alpha c}$$

$$\text{MLE for } \hat{\theta}_{\alpha c} = \frac{|\{i | x_i^\alpha = 1, y_i = c\}|}{|\{i | y_i = c\}|}$$

also let  ~~$\hat{\theta}(y)$~~  =  $\frac{|\{i | y_i = c\}|}{n}$   
 $\hat{P}(y=c)$

$\{x | f(x)\} \equiv$  set of  $x$  such that  $f(x)$  is true.

$|A| =$  size of set  $A$

$$\frac{\sum_{i=1}^n \mathbb{I}(x_i^\alpha = 1 \text{ and } y_i = c)}{\sum_{i=1}^n \mathbb{I}(y_i = c)}$$

$$\sum_{i=1}^n \mathbb{I}(y_i = c)$$