# 朴素贝叶斯

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https://simplelp.github.io/

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### 朴素贝叶斯分类器

$$y = f(x) = \arg \max_{c_k} \frac{P(Y = c_k) \prod_{j} P(X^{(j)} = x^{(j)} | Y = c_k)}{\sum_{k} P(Y = c_k) \prod_{j} P(X^{(j)} = x^{(j)} | Y = c_k)}$$

$$y = \arg \max_{c_k} P(Y = c_k) \prod_{j} P(X^{(j)} = x^{(j)} | Y = c_k)$$

### 参数估计: 最大似然估计

$$P(Y = c_k) = \frac{\sum_{i=1}^{N} I(y_i = c_k)}{N}, \quad k = 1, 2, \dots, K$$

$$P(X^{(j)} = a_{jl} \mid Y = c_k) = \frac{\sum_{i=1}^{N} I(x_i^{(j)} = a_{jl}, y_i = c_k)}{\sum_{i=1}^{N} I(y_i = c_k)}$$

$$j = 1, 2, \dots, n$$
;  $l = 1, 2, \dots, S_j$ ;  $k = 1, 2, \dots, K$ 

## 参数估计: 贝叶斯估计

$$P_{\lambda}(Y=c_{k}) = \frac{\sum_{i=1}^{N} I(y_{i}=c_{k}) + \lambda}{N + K\lambda}$$

$$P_{\lambda}(X^{(j)} = a_{jl} \mid Y = c_{k}) = \frac{\sum_{i=1}^{N} I(x_{i}^{(j)} = a_{jl}, y_{i} = c_{k}) + \lambda}{\sum_{i=1}^{N} I(y_{i} = c_{k}) + S_{j}\lambda}$$

#### 思考

- •证明后验概率最大化等价于期望风险最小化(可参考课本)
- 两种参数估计的证明
- 利用两种参数估计方法对例题4.1进行分类
- 代码实现朴素贝叶斯分类器

# Thanks!