

# MSAI Probability Week 2

- Conditional Probability
- Maximum Likelihood Estimate
- Naïve Bayes Classifier

P2) 1)  $p(B|A) + p(B|\bar{A}) \stackrel{?}{=} 1 \leftarrow \text{False}$

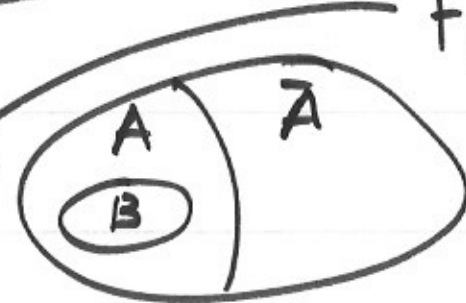
$$p(B|A)p(A) + p(B|\bar{A})p(\bar{A}) = p(B)$$
$$p(A, B) + p(B, \bar{A}) = p(B)$$

Pick  $B = \Omega$

$$p(\Omega|A) + p(\Omega|\bar{A}) = 2$$

False

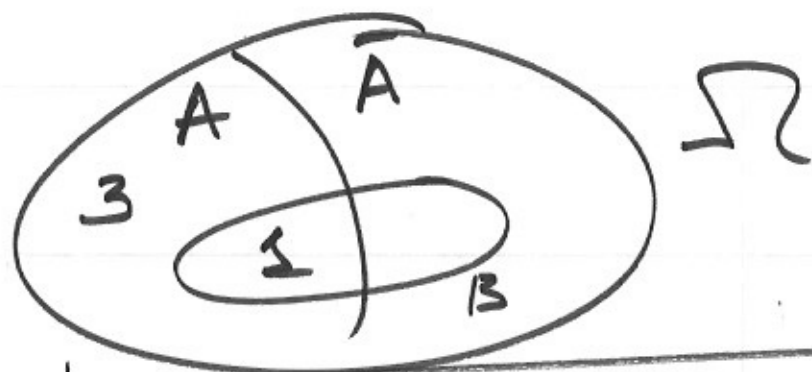
2)  $p(B|A) + p(\bar{B}|\bar{A}) = 1$



$B \subset A \rightarrow p(\bar{B}|\bar{A}) = p(\bar{A})$

Pick  $B = A \rightarrow p(A|A) + p(\bar{A}|\bar{A}) = 2$

⑤



$$\boxed{P(B|A) + P(\bar{B}|A) = 1}$$

$$\frac{1}{3+1} \quad \frac{3}{3+1}$$

Bayes' formula

$$P(A, B) = \boxed{P(A|B) \cdot P(B) = P(B|A) \cdot P(A)}$$

$$P(A|B) = \frac{P(B|A) \cdot P(A)}{P(B)}$$

(2)

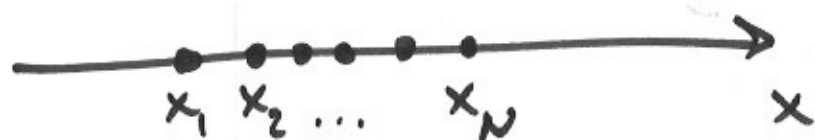
# Maximum Likelihood Estimation



$$x \sim \mathcal{N}(\mu, \sigma)$$

$$p(x | \mu, \sigma) = \frac{1}{\sqrt{2\pi\sigma^2}} \cdot \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right)$$

$(2\pi\sigma^2)^{-1/2}$



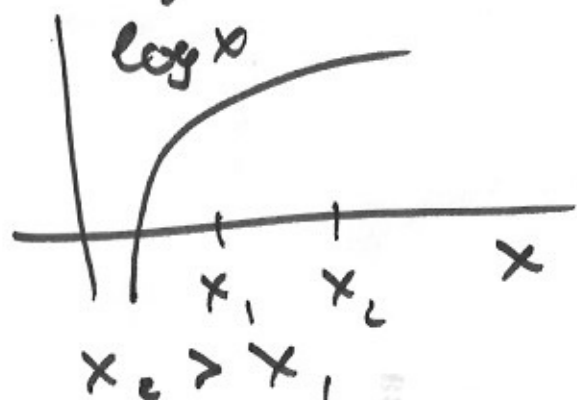
assume

$\tilde{\mu}, \tilde{\sigma}$

$$p(\mu, \sigma | \vec{x} = (x_1, x_2, \dots)) = \frac{p(\vec{x} | \mu, \sigma) \cdot p(\mu, \sigma)}{p(\vec{x})}$$

posterior =  $\frac{\text{likelihood} \cdot \text{prior}}{\text{evidence}}$   
(of  $\mu, \sigma$ )

$$\arg \max_{\mu, \sigma} L(\mu, \sigma) = \arg \max_{\mu, \sigma} \log L(\mu, \sigma)$$



$$\log \prod_i = \sum_i \log$$

$$\rightarrow \log x_2 > \log x_1$$

$$\log L = \sum_{i=1}^N \log \left( (2\pi\sigma^2)^{-1/2} \exp\left(-\frac{(x_i - \mu)^2}{2\sigma^2}\right) \right)$$

$$= -\frac{N}{2} \log(2\pi\sigma^2) - \sum_{i=1}^N \frac{1}{2\sigma^2} (x_i - \mu)^2$$

$$\frac{\partial}{\partial \mu} (\dots) = 0 - \sum \frac{1}{2\sigma^2} \cdot 2(x_i - \mu) \cdot (-1) =$$

$$= \sum \frac{1}{\sigma^2} (x_i - \mu) = 0 \rightarrow \sum x_i = \mu \cdot N$$

$$\dots \rightarrow \sigma = \sqrt{\frac{1}{N} \sum (x_i - \mu)^2}$$

$$\tilde{\mu} = \frac{1}{N} \sum x_i$$

④