

## Density Estimation

Many approaches make assumptions about the probability distribution of our data.

- naïve Bayes

$$p(x_1, x_2, \dots, x_D | \omega) = \prod_{i=1}^D p(x_i | \omega)$$

## Parametric Model

These are models

$$p(X=x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

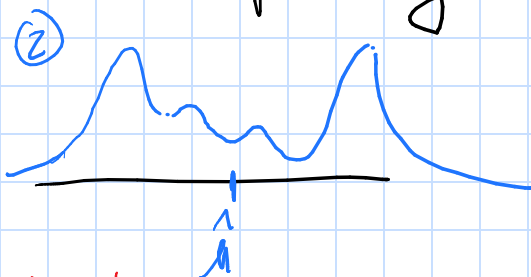
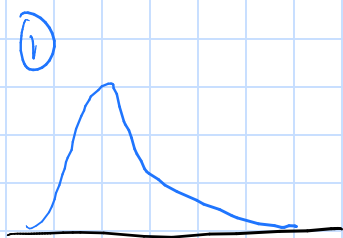
Parameters:  $\mu, \sigma$

$$p(y|x) = \frac{p(x|y)p(y)}{p(x)} \rightarrow \begin{array}{l} \text{Easy to estimate} \\ \text{Total probability Thm} \end{array}$$

(1) How do we find the term?

(2) How good of a fit is our model to the data?

- KS - test



## Maximum Likelihood Estimator

$$\hat{\theta} = \arg \max_{\theta \in \Phi} \{p(D|\theta)\} = \arg \max_{\theta \in \Phi} \{\log p(D|\theta)\}$$

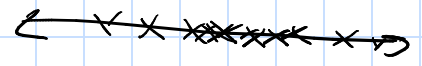
- Choose the  $\hat{\theta}$  that best describes the data.

- Why do we take the log probability?

$$p(D|\theta) = \prod_{i=1}^n p(x_i|\theta) \Rightarrow \log p(D|\theta) = \sum_{i=1}^n \log p(x_i|\theta)$$

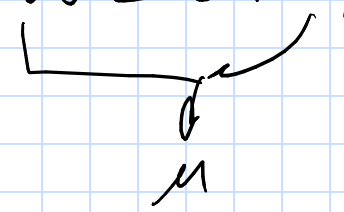
Maximize  $\ln L$

$$\nabla_{\theta} L(\theta) = \sum_{i=1}^n \nabla_{\theta} \log p(x_i | \theta) = 0$$

  
 $n = 75$

ex/ Gaussian distribution  $\theta = \mu$ ,  $\Sigma$  is known

$$\log p(x_i | \theta) = -\frac{1}{2} \log \{ (2\pi)^d |\Sigma| \} - \frac{1}{2} (x_i - \theta)^T \Sigma^{-1} (x_i - \theta)$$

  
 $\mu$

$$\nabla_{\theta} \log p(x_i | \theta) = 0 = \Sigma^{-1} (x_i - \theta)$$

$$\sum_{i=1}^n \Sigma^{-1} (x_i - \theta) = 0$$

$$\Sigma^{-1} \left( \sum_{i=1}^n (x_i - \theta) \right) = 0$$

$$\text{then } \theta = \frac{1}{n} \sum_{i=1}^n x_i$$

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### Bayesian Estimation

$$\theta_{BE} = \arg \max \{ p(\theta | D) \} = \arg \max \{ p(\theta) p(D | \theta) \}$$

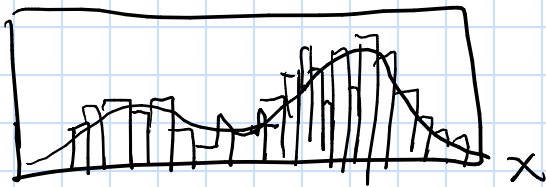
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### Nonparametric Density Estimation

Problems w/ parametric models

- Assume the form of the probability distribution
- Most distributions are unimodal

# Density Estimation 101



$$P_i = \frac{n_i}{N \Delta_i}$$

$$p = \int_R p(x') dx'$$

↳ smoothed or averaged version of  $p(x)$

$$\int_R p(x') dx' \approx p(x) \cdot V$$

$x \rightarrow$  point in  $R$

$V \rightarrow$  volume enclosed by the region  $R$

$k \rightarrow$  # of samples that fall in a region  $R$

$$\Rightarrow \boxed{p_n(x) = \frac{k/n}{V}}$$

$$\frac{E[k]}{V} = n p(x)$$

$$V \rightarrow 0 ?$$

$$n \rightarrow \infty ?$$