

Model based approach

1 Our goal is to build parametric model for conditional distribution $P(Y=k|X=x)$

$$P_r(Y=k|X=x) = \frac{P_r(X=x|Y=k) P_r(Y=k)}{\sum_{\ell=1}^K P_r(X=x|Y=\ell) P_r(Y=\ell)}$$

$$P_r(Y=k|X=x) = \frac{f_k(x) \pi_k}{\sum_{\ell=1}^K f_{\ell}(x) \pi_{\ell}}$$

π_k prior probability

A common choice $f_k(x) = \frac{1}{\sigma_k \sqrt{2\pi}} e^{-\frac{(x-\mu_k)^2}{\sigma_k^2}}$

Estimate the parameters (μ_k, σ_k^2) from the data

Assigning to the class which has the highest value of $P(Y=k|X=x)$

$$\log \frac{P_r(Y=k|X=x)}{P_r(Y=j|X=x)} =$$

$$= \log \frac{f_k(x)}{f_j(x)} + \log \frac{\sigma_k}{\sigma_j} =$$

$$= \log \frac{\sigma_k}{\sigma_j} - \frac{1}{2} (\mu_k + \mu_j)^T \Sigma^{-1} (\mu_k + \mu_j) +$$

$$+ x^T \Sigma^{-1} (\mu_k - \mu_j)$$

Discriminant function

$$\delta_k(x) = x^T \Sigma^{-1} \mu_k - \frac{1}{2} \mu_k^T \Sigma^{-1} \mu_k + \log(\mu_k)$$

- Decide on class based on $\hat{y}(x) = \arg \max_k \delta_k(x)$
- we usually do this with EM.