

AIC & BIC

Mixture Models

Bayesian & conj. prior

~~Double descent~~

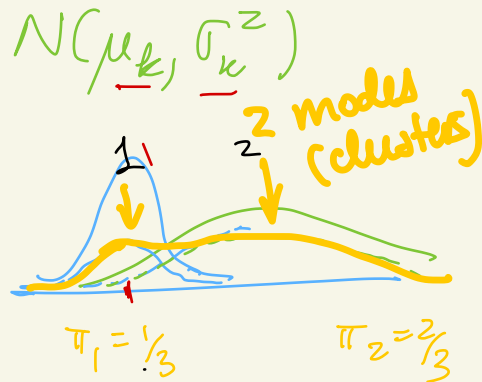
Mixture model

f = weighted sum of K densities

$$f(x) = \sum \pi_k f_k(x)$$

≥ 0 for all x ✓ density

$$\int_{-\infty}^{\infty} f(x) dx = 1 \quad \checkmark$$



$x \sim f(x)$: sample $k \in 1:K \sim \underline{\pi}_{1:K}$
sample $x \sim N(\underline{\mu}_k, \underline{\sigma}_k^2) = f_k$
Generative model

parameters:

$$x \in \mathbb{R}^d$$

$$\mu_{1:k}$$

$$k$$

$$kd$$

$$\Sigma_{1:k}^2$$

$$k$$

$$k \frac{d(d+1)}{2}$$

$$\Pi_{1:k}$$

$$k-1$$

$$k-1$$

$$\Sigma \bar{u}_k = \perp$$

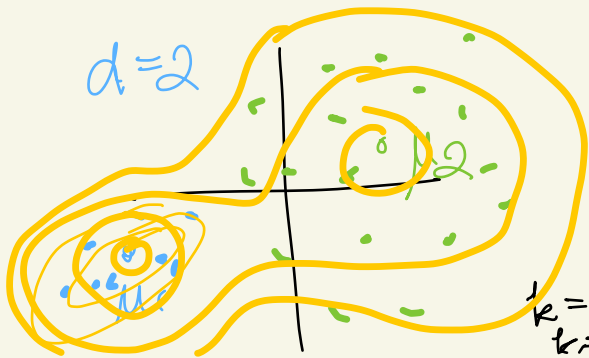
$$3k-1$$

$$\in \mathbb{R}^2$$

$$N_{1,2,1}$$

$$\Sigma_{1,2} = \begin{bmatrix} \Sigma_1^2 & \Sigma_{1,2} \\ \Sigma_{1,2} & \Sigma_2^2 \end{bmatrix}$$

$$d=2$$



$$k=1$$

$$d=1$$

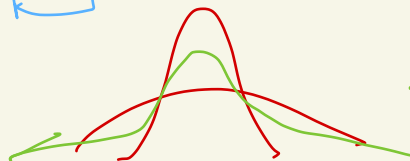
$$n_1=5$$

$$n_2=11$$



$$k=2$$

$$\boxed{\Sigma}$$

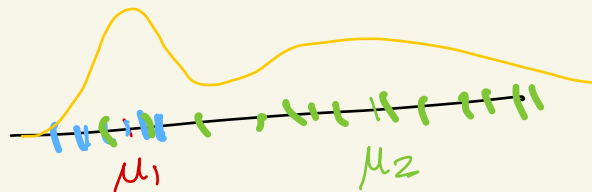


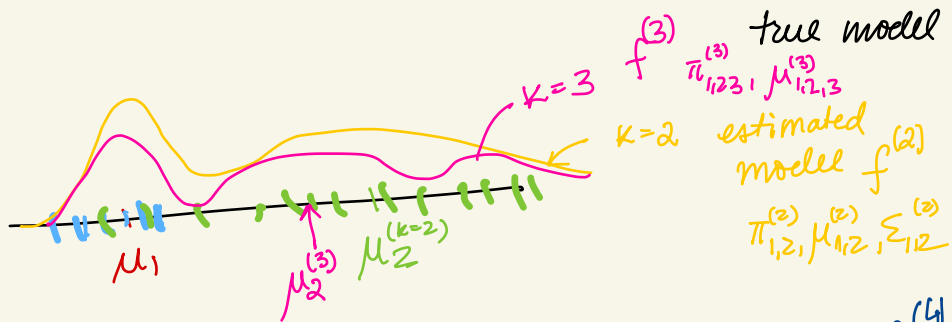
← symmetric



$$d^2 - \frac{d(d-1)}{2}$$

(free) parameters





$f^{(3)}$ true model
 $\pi_{1,2,3}^{(3)}, \mu_{1,2,3}^{(3)}$
 $k=2$ estimated model $f^{(2)}$
 $\pi_{1,2}^{(2)}, \mu_{1,2}^{(2)}, \Sigma_{1,2}^{(2)}$

$k=2$
 $\Rightarrow$ params: $3 + 3 + 2$
 $\mu \quad \sigma^2 \quad \tau_0$

$\# \text{params} = 2 + 2 + 1$
 $\mu \quad \sigma^2 \quad \tau$
 $4 + 4 + 3$

$k=4 \Rightarrow f^{(4)}$

likelihood

$\ell(\mathcal{D} | f^{(4)}) \geq \ell(\mathcal{D} | f^{(3)}) \geq \ell(\mathcal{D} | f^{(2)})$

$$AIC(f^{(2)}) = \ell(\mathcal{D} | f^{(2)}) - \underbrace{5}_{\# \text{params}}$$

$$AIC(f^{(3)}) = \ell(\mathcal{D} | f^{(3)}) - (3 + 3 + 2)$$

$$AIC(f^{(4)}) = \ell(\mathcal{D} | f^{(4)}) - (4 + 4 + 3)$$

choose max

$$BIC(f^{(4)}) = \ell(\mathcal{D} | f^{(4)}) - \underbrace{\left[\frac{\ln n}{2} \right]}_{\substack{\text{ML from} \\ \text{model} \\ \text{class}}} \underbrace{(4 + 4 + 3)}_{\text{nr. parameters}}$$

only parametric (no KDE)

(Parametric)
Model family

{Normal}

estimation
 $\xrightarrow{\text{ML}}$

$$f^{\text{ML}} \equiv \mu^{\text{ML}}, (\Sigma^z)^{\text{ML}}$$

{Logistic}

Select model

CV

AIC, BIC

{Poisson}

...

{Mixture $k=2$ Normal} $\rightarrow f^{\text{ML}} = \mu_{1,2}^{\text{ML}}, \Sigma_{1,2}^z, \pi_{1,2}^{\text{ML}}$

{ " " $k=3$ " " }

...

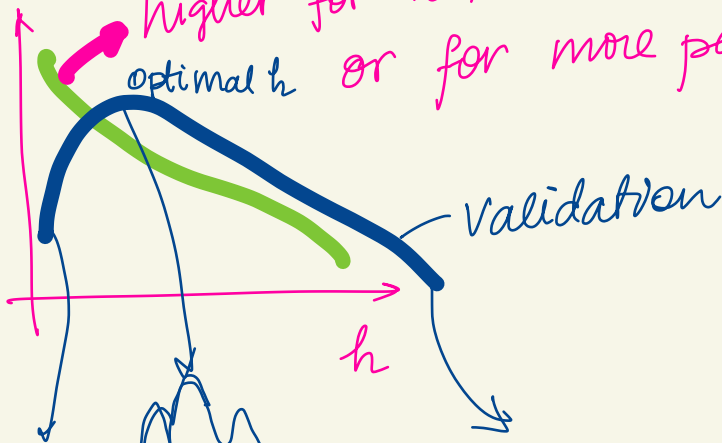
{ " " $k=3$ Logistic }

Nonparametric KDE { $f_{h,K(\cdot),n}$ } $\xrightarrow{\text{KDE}}$ f

Select
model

CV

$\log^{-1}$
higher for h smaller
optimal h or for more parameters



High var
Low bias
training set

Bias
Low var