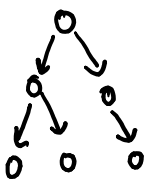


MAP Estimation

Thursday, June 8, 2017 10:03 AM

DAG - directed acyclic graph

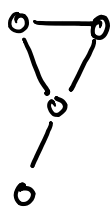


DAG conditional indep is $X_i \perp\!\!\!\perp X_k \mid X_{P_i}$ if P_i is parent of i
 k is not a descendant of i .

def The moral graph of a DAG has an undirected edge btw X_i, X_j if

(i) \exists a directed edge btw X_j, X_i or

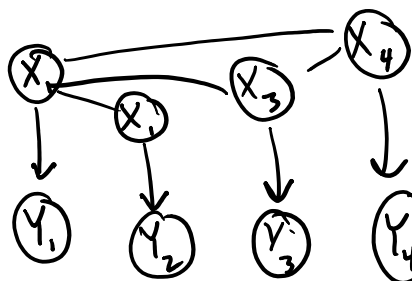
(ii) X_j, X_i are parents of the same node



DAG

\Rightarrow Moral graph has global markov property

Hidden Markov Models



Latent RVs

Observables

use cond. density $1_{X|Y}(x|y)$ to estimate X



X is w/in discrete space



Maximum A-posteriori estimator (MAP)

$$\hat{x} = \underset{x}{\operatorname{argmax}} p_{X|Y}(x|y) = \underset{x}{\operatorname{argmax}} p_X(x) \cdot p_{Y|X}(y|x)$$

$$p_{X|Y}(x|y) = \frac{p_{X,Y}(x,y)}{\underbrace{p_Y(y)}_{\text{constant in } x}}$$

$$-\log p_{Y|X}(y|x) = \frac{1}{n} \sum_{i=1}^n \ell(y_i|x_i)$$

$$-\log p_X(x) = \rho(x)$$

$$\text{MAP: } \min_x \frac{1}{n} \sum_{i=1}^n \ell(y_i|x_i) + \rho(x)$$

Ising model for $X \in \{-1, 1\}^n$

$$p_X(x) = \frac{1}{Z} \exp \left\{ \sum_{i,j} w_{ij} x_i x_j + \sum_i a_i x_i \right\}$$

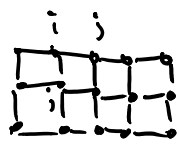
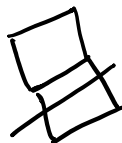


image $p(x) \propto \sum_{i,j} w_{ij} x_i x_j + \sum_i a_i x_i$

▷ for the Binary case this is solvable as a graph cut problem.



cut = 1

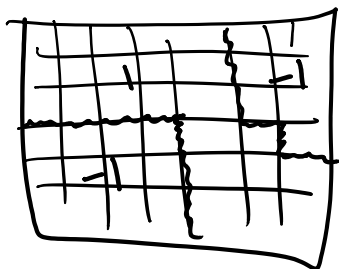


cut = 2

MAP for Ising

$$\min_{x \in \{-1, 1\}^n} \frac{1}{n} \sum_i \ell(y_i|x_i) + \sum_{i,j} w_{ij} x_i x_j + \sum_i a_i x_i$$

$$= \sum_i b_i x_i + \sum_{i \sim j} w_{ij} x_i x_j$$



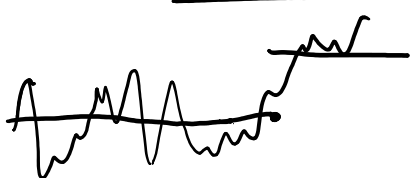
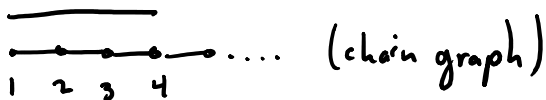
Total Variation denoising

$$\min_{x \in \mathbb{R}^n} \frac{1}{n} \sum_{i=1}^n \ell(y_i | x_i) + \lambda \sum_{i \sim j} w_{ij} |x_i - x_j|$$

aka fused lasso

for square error loss: $\min_x \frac{1}{n} \sum_{i=1}^n (y_i - x_i)^2 + \lambda \sum_{i \sim j} w_{ij} |x_i - x_j|$

can be used for time series ...



$$w_{ij} = \begin{cases} 1, & |i-j|=1 \\ 0, & \text{otherwise} \end{cases}$$

