

# Recap

## Undirected graphical models

- ▶ graph  $G$ :  
set of vertices/nodes  $V = \{1, \dots, p\}$   
set of edges  $E \subseteq V \times V$
- ▶ random variables  $X = X^{(1)}, \dots, X^{(p)}$  with distribution  $P$   
identify nodes in  $V$  with components of  $X$

graphical model:  $(G, P)$

pairwise Markov property:

$P$  satisfies the pairwise Markov property (w.r.t.  $G$ ) if

$$(j, k) \notin E \implies X^{(j)} \perp X^{(k)} \mid X^{(V \setminus \{j, k\})}$$

## Global Markov property

(stronger property than pairwise Markov prop):

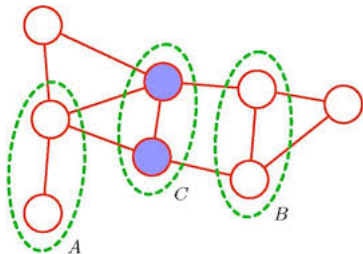
consider disjoint subsets  $A, B, C \subseteq V$

$P$  satisfies the global Markov property (w.r.t.  $G$ ) if

$A$  and  $B$  are separated by  $C \implies X^{(A)} \perp X^{(B)} \mid$

$X^{(C)}$

only condition on subset  $C$



global Markov property  $\implies$  pairwise Markov property

if  $P$  has a positive density w.r.t Lebesgue measure:  
global Markov property  $\iff$  pairwise Markov property

Gaussian Graphical Model (GGM):  $P \sim \mathcal{N}_p(\mathbf{0}, \Sigma)$ :

$$(j, k) \in E \iff (\Sigma^{-1})_{jk} \neq 0$$

## Neighborhood selection: nodewise regression

$$X^{(j)} = \beta_k^{(j)} X^{(k)} + \sum_{r \neq j, k} \beta_r^{(j)} X^{(r)} + \varepsilon^{(j)}, \quad j = 1, \dots, p$$

$$X^{(k)} = \beta_j^{(k)} X^{(j)} + \sum_{r \neq k, j} \beta_r^{(k)} X^{(r)} + \varepsilon^{(k)}$$

for GGM:

$$(j, k) \in E \iff \beta_k^{(j)} \neq 0 \iff \beta_j^{(k)} \neq 0$$

nodewise regression (Meinshausen & Bühlmann, 2006)

- ▶ run Lasso for every node variable  $X^{(j)}$  versus all others  $\{X^{(k)}; k \neq j\}$  ( $j = 1, \dots, p$ )
- ▶ estimated active set  $\hat{S}^{(j)} = \{r; \hat{\beta}_r^{(j)} \neq 0\}$  ( $j = 1, \dots, p$ )
- ▶ estimate edges in  $\hat{E}$  :

or rule:  $(j, k) \in \hat{E} \iff j \in \hat{S}^{(k)}$  **or**  $k \in \hat{S}^{(j)}$

and rule:  $(j, k) \in \hat{E} \iff j \in \hat{S}^{(k)}$  **and**  $k \in \hat{S}^{(j)}$

just run Lasso  $p$  times: it's fast!

R-packages `huge` and also in `glasso` (and set 'approx = T')