GRAPHICAL MODELS

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6. DECOMPOSABLE GRAPHS (triangulated graphs, chordal graphs)

A part of this chapter is based on lectures of Prof. S. Lauritzen at CIMPA Summer School Hammamet 2011, with his kind permission.

Only graphical models governed by **DECOMPOSABLE GRAPHS** have good statistical properties:

- one can compute easily MLE estimators \hat{K} and $\hat{\Sigma}$ of the precision and covariance matrices
- statistical tests can be performed
- Bayesian statistics is possible and performant

That's why we shall learn some theory of DECOMPOSABLE GRAPHS

Consider an undirected graph $\mathcal{G} = (V, E)$ with vertices V and edges E.

If $W \subset V$, the **induced graph** is $\mathcal{G}_W = (W, E_W)$ where $\{i, j\} \in E_W$ if and only if $\{i, j\} \in E$ and $i, j \in W$. The edges of the **induced graph** \mathcal{G}_W are all the edges of \mathcal{G} connecting vertices from W.

A path of length *n* from $\alpha \in V$ to $\beta \in V$ is a sequence

$$\alpha_0 = \alpha, \alpha_1, \dots, \alpha_n = \beta$$

of vertices distinct for i = 0, ..., n - 1 such that $\{\alpha_i, \alpha_{i+1}\} \in E$ for each i = 1, ..., n.

A subset $S \subset V$ is an (α, β) -**separator** if every path from α to β intersects S.

S separates $A \subset V$ from $B \subset V$ if S is an (α, β) separator for every $\alpha \in A$ and $\beta \in B$.

A separator of A and B is **minimal** if no proper subset $T \subsetneq S$ separates A and B.

A graph is **complete** if all vertices are joined by an edge. A subset W is complete if its induced graph \mathcal{G}_W is complete.

A clique of \mathcal{G} is a maximal complete subset of V.

A cycle of length n is a **path** of length n from α to α . The shortest cycles are triangles=cycles of length 3.

A **tree** is a connected graph without cycles. It has a unique path between any two vertices.

A graph is triangulated(chordal) if every cycle of length $n \ge 4$ has a chord, that is two non-consecutive vertices that are connected by an edge(chord).





the set $S = \{1, 3\}$ is a (2, 4)-separator. The separator S is minimal. S is not complete.

the set $S' = \{2, 4\}$ is a (1, 3)-separator. The separator S' is minimal. S' is not complete.

There are no other separators. No separator is complete.

The cliques are $\{1,2\}$, $\{2,3\}$, $\{3,4\}$ and $\{1,4\}$



the set $S = \{1,3\}$ is a (2,4)-separator. S is minimal and complete. There are no other separators. (the set $S' = \{2,4\}$ is NOT a (1,3)-separator)

Every minimal separator is complete.

The cliques are $\{1, 2, 3\}$ and $\{1, 3, 4\}$.

Definition Decomposition of Markov properties Factorization of Markov distributions Properties of decomposability

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Consider an *undirected* graph $\mathcal{G} = (V, E)$. A partitioning of V into a triple (A, B, S) of subsets of V forms a *decomposition* of \mathcal{G} if

 $A \perp_{\mathcal{G}} B \mid S$ and S is complete.

The decomposition is *proper* if $A \neq \emptyset$ and $B \neq \emptyset$. The *components* of \mathcal{G} are the induced subgraphs $\mathcal{G}_{A\cup S}$ and $\mathcal{G}_{B\cup S}$. A graph is *prime* if no proper decomposition exists.

Definition Decomposition of Markov properties Factorization of Markov distributions Properties of decomposability

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Examples



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Suppose P satisfies (F) w.r.t. G and (A, B, S) is a decomposition. Then

(i) $P_{A\cup S}$ and $P_{B\cup S}$ satisfy (F) w.r.t. $\mathcal{G}_{A\cup S}$ and $\mathcal{G}_{B\cup S}$ respectively; (ii) $f(x)f_S(x_S) = f_{A\cup S}(x_{A\cup S})f_{B\cup S}(x_{B\cup S})$.

The converse also holds in the sense that *if* (*i*) and (*ii*) hold, and (A, B, S) is a decomposition of \mathcal{G} , then P factorizes w.r.t. \mathcal{G} .

Definition Decomposition of Markov properties Factorization of Markov distributions Properties of decomposability

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Decomposability

Any graph can be recursively decomposed into its maximal prime subgraphs: 2 4



A graph is *decomposable* (or rather fully decomposable) if it is complete or admits a proper decomposition into *decomposable* subgraphs.

Definition is recursive. Alternatively this means that *all maximal prime subgraphs are cliques*.

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Recursive decomposition of a decomposable graph into cliques yields the formula:

$$f(x)\prod_{S\in\mathcal{S}}f_S(x_S)^{\nu(S)}=\prod_{C\in\mathcal{C}}f_C(x_C).$$

Here S is the set of *minimal complete separators* occurring in the decomposition process and $\nu(S)$ the number of times such a separator appears in this process.

Definition Decomposition of Markov properties Factorization of Markov distributions Properties of decomposability

Perfect numbering

A numbering $V = \{1, ..., |V|\}$ of the vertices of an undirected graph is *perfect* if

 $\forall j = 2, \ldots, |V| : \mathsf{bd}(j) \cap \{1, \ldots, j-1\}$ is complete in \mathcal{G} .

A set S is an (α, β) -separator if $\alpha \perp_{\mathcal{G}} \beta \mid S$,

Definition Decomposition of Markov properties Factorization of Markov distributions Properties of decomposability

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Characterizing chordal graphs

The following are equivalent for any undirected graph \mathcal{G} .

- (i) \mathcal{G} is chordal;
- (ii) \mathcal{G} is decomposable;
- (iii) All maximal prime subgraphs of \mathcal{G} are cliques;
- (iv) \mathcal{G} admits a perfect numbering;
- (v) Every minimal (α, β) -separator are complete.

Trees are chordal graphs and thus decomposable.

Here is a (greedy) algorithm for checking chordality:

- 1. Look for a vertex v^* with $bd(v^*)$ complete. If no such vertex exists, the graph is not chordal.
- 2. Form the subgraph $\mathcal{G}_{V\setminus v^*}$ and let $v^* = |V|$;
- 3. Repeat the process under 1;
- 4. If the algorithm continues until only one vertex is left, the graph is chordal and the numbering is perfect.

The complexity of this algorithm is $O(|V|^2)$.

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This graph is *not* chordal, as there is no candidate for number 4.

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This graph is chordal!

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This simple algorithm has complexity O(|V| + |E|):

- 1. Choose $v_0 \in V$ arbitrary and let $v_0 = 1$;
- 2. When vertices $\{1, 2, \ldots, j\}$ have been identified, choose v = j + 1 among $V \setminus \{1, 2, \ldots, j\}$ with highest cardinality of its numbered neighbours;
- 3. If $bd(j+1) \cap \{1, 2, \dots, j\}$ is not complete, \mathcal{G} is not chordal;
- 4. Repeat from 2;
- 5. If the algorithm continues until no vertex is left, the graph is chordal and the numbering is perfect.

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The graph is not chordal! because 7 does not have a complete boundary.

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Maximum Cardinality Search



MCS numbering for the chordal graph. Algorithm runs essentially as before.

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A chordal graph



This graph is chordal, but it might not be that easy to see... Maximum Cardinality Search is handy!

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Greedy algorithm Maximum cardinality search

Finding the cliques of a chordal graph

From an MCS numbering $V = \{1, \ldots, |V|\}$, let

$$B_\lambda = \mathsf{bd}(\lambda) \cap \{1, \dots, \lambda - 1\}$$

and $\pi_{\lambda} = |B_{\lambda}|$. Call λ a *ladder vertex* if $\lambda = |V|$ or if $\pi_{\lambda+1} < \pi_{\lambda} + 1$. Let Λ be the set of ladder vertices.



 π_{λ} : 0,1,2,2,2,1,1. The cliques are $C_{\lambda} = \{\lambda\} \cup B_{\lambda}, \lambda \in \Lambda$.

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Definition Characterizing chordal graphs Construction of junction tree Junction trees of prime components

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Let \mathcal{A} be a collection of finite subsets of a set V. A *junction tree* \mathcal{T} of sets in \mathcal{A} is an undirected tree with \mathcal{A} as a vertex set, satisfying the *junction tree property:*

If $A, B \in \mathcal{A}$ and C is on the unique path in \mathcal{T} between A and B it holds that $A \cap B \subset C$.

If the sets in an arbitrary \mathcal{A} are pairwise incomparable, *they can be arranged in a junction tree if and only if* $\mathcal{A} = \mathcal{C}$ *where* \mathcal{C} *are the cliques of a chordal graph*

Definition Characterizing chordal graphs Construction of junction tree Junction trees of prime components

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The following are equivalent for any undirected graph \mathcal{G} .

- (i) \mathcal{G} is chordal;
- (ii) G is decomposable;
- (iii) All prime components of \mathcal{G} are cliques;
- (iv) *G* admits a perfect numbering;
- (v) Every minimal (α, β) -separator are complete.
- (vi) The cliques of \mathcal{G} can be arranged in a junction tree.

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The junction tree can be *constructed directly from the MCS* ordering $C_{\lambda}, \lambda \in \Lambda$, where C_{λ} are the cliques: Since the MCS-numbering is perfect, $C_{\lambda}, \lambda > \lambda_{\min}$ all satisfy

$$C_{\lambda} \cap (\cup_{\lambda' < \lambda} C_{\lambda'}) = C_{\lambda} \cap C_{\lambda^*} = S_{\lambda}$$

for some $\lambda^* < \lambda$.

A junction tree is now easily constructed by attaching C_{λ} to any C_{λ^*} satisfying the above. Although λ^* may not be uniquely determined, S_{λ} is.

Indeed, the sets S_{λ} are the minimal complete separators and *the* numbers $\nu(S)$ are $\nu(S) = |\{\lambda \in \Lambda : S_{\lambda} = S\}|$.

Definition Characterizing chordal graphs Construction of junction tree Junction trees of prime components

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A chordal graph



Definition Characterizing chordal graphs Construction of junction tree Junction trees of prime components

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Junction tree



Cliques of graph arranged into a tree with $C_1 \cap C_2 \subseteq D$ for all cliques D on path between C_1 and C_2 .

Definition Characterizing chordal graphs Construction of junction tree Junction trees of prime components

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In general, the *prime components* of any undirected graph can be arranged in a junction tree in a similar way.

Then every pair of neighbours (C, D) in the junction tree represents a decomposition of \mathcal{G} into $\mathcal{G}_{\tilde{C}}$ and $\mathcal{G}_{\tilde{D}}$, where \tilde{C} is the set of vertices in prime components connected to C but separated from D in the junction tree, and similarly with \tilde{D} .

The corresponding algorithm is based on a slightly more sophisticated algorithm known as *Lexicographic Search* (LEX) which runs in $O(|V|^2)$ time.

Basic factorizations Maximum likelihood estimates An example

If the graph \mathcal{G} is chordal, we say that the graphical model is *decomposable*.

In this case, the IPS-algorithm converges in a finite number of steps.

We also have the familiar *factorization of densities*

$$f(x \mid \Sigma) = \frac{\prod_{C \in \mathcal{C}} f(x_C \mid \Sigma_C)}{\prod_{S \in \mathcal{S}} f(x_S \mid \Sigma_S)^{\nu(S)}}$$
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where $\nu(S)$ is the number of times S appear as intersection between neighbouring cliques of a junction tree for C.

Basic factorizations Maximum likelihood estimates An example

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Relations for trace and determinant

Using the factorization (1) we can for example match the expressions for the trace and determinant of $\boldsymbol{\Sigma}$

$$\operatorname{tr}(KW) = \sum_{C \in \mathcal{C}} \operatorname{tr}(K_C W_C) - \sum_{S \in \mathcal{S}} \nu(S) \operatorname{tr}(K_S W_S)$$

and further

$$\det \Sigma = \{\det(K)\}^{-1} = \frac{\prod_{C \in \mathcal{C}} \det\{\Sigma_C\}}{\prod_{S \in \mathcal{S}} \{\det(\Sigma_S)\}^{\nu(S)}}$$

These are some of many relations that can be derived using the decomposition property of chordal graphs.

Basic factorizations Maximum likelihood estimates An example

The same factorization clearly holds for the maximum likelihood estimates:

$$f(x \mid \hat{\Sigma}) = \frac{\prod_{C \in \mathcal{C}} f(x_C \mid \hat{\Sigma}_C)}{\prod_{S \in \mathcal{S}} f(x_S \mid \hat{\Sigma}_S)^{\nu(S)}}$$
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Moreover, it follows from the general likelihood equations that

 $\hat{\Sigma}_{A} = W_{A}/n$ whenever A is complete.

Exploiting this, we can obtain an explicit formula for the maximum likelihood estimate in the case of a chordal graph.

Basic factorizations Maximum likelihood estimates An example

For a $|d| \times |e|$ matrix $A = \{a_{\gamma\mu}\}_{\gamma \in d, \mu \in e}$ we let $[A]^V$ denote the matrix obtained from A by filling up with zero entries to obtain full dimension $|V| \times |V|$, i.e.

$$([A]^V)_{\gamma\mu} = \begin{cases} a_{\gamma\mu} & \text{if } \gamma \in d, \mu \in e \\ 0 & \text{otherwise.} \end{cases}$$

The maximum likelihood estimates exists if and only if $n \ge C$ for all $C \in C$. Then the following simple formula holds for the maximum likelihood estimate of K:

$$\hat{K} = n \left\{ \sum_{C \in \mathcal{C}} \left[(w_C)^{-1} \right]^V - \sum_{S \in \mathcal{S}} \nu(S) \left[(w_S)^{-1} \right]^V \right\}$$

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Suppose that the graph G is decomposable. Let Cliq be the set of all cliques of G and Sep the set of all minimal separators of G.

Suppose that $n \ge |C|$ (the number of elements of *C*) for each clique *C*.

If the mean ξ of the model is known and $\tilde{\Sigma}$ is the sample covariance matrix then

$$\hat{K} = \sum_{C \in Cliq} [\tilde{\Sigma}_C^{-1}]^V - \sum_{S \in Sep} \nu(S) [\tilde{\Sigma}_S^{-1}]^V$$

If the mean is unknown, then $\hat{\xi} = \bar{X}$ and one uses the "Clique-separator formula" for \hat{K} with the corrected sample covariance matrix $\frac{n}{n-1}\tilde{\Sigma}$. Back to Example "Simpson paradox" \mathcal{G} : 1—3—2 Suppose that $\xi = 0$ and the sample covariance matrix equals

$$\tilde{\Sigma} = \begin{pmatrix} 1 & 0.5 & 1 \\ 0.5 & 2 & 2 \\ 1 & 2 & 3 \end{pmatrix}$$
. The graph \mathcal{G} governs the model.
We computed "by hand" $\hat{\Sigma} = \begin{pmatrix} 1 & \frac{2}{3} & 1 \\ \frac{2}{3} & 2 & 2 \\ 1 & 2 & 3 \end{pmatrix}$

Let us find \hat{K} and $\hat{\Sigma}$ by "Clique-separator formula".

The cliques of \mathcal{G} are $C_1 = \{1,3\}$ and $C_2 = \{2,3\}$. The minimal separator is $S = \{3\}$.

$$\begin{split} \tilde{\Sigma} &= \begin{pmatrix} 1 & 0.5 & 1 \\ 0.5 & 2 & 2 \\ 1 & 2 & 3 \end{pmatrix}. \text{ We only use } \pi_{\mathcal{G}}(\tilde{\Sigma}) = \begin{pmatrix} 1 & 1 \\ 2 & 2 \\ 1 & 2 & 3 \end{pmatrix} \\ \text{Apply the "Clique-separator formula" for } \tilde{K}: \\ \tilde{K} &= [\tilde{\Sigma}_{1,3}^{-1}]^V + [\tilde{\Sigma}_{2,3}^{-1}]^V - [\tilde{\Sigma}_{3}^{-1}]^V. \\ \tilde{\Sigma}_{1,3}^{-1} &= \begin{pmatrix} 1 & 1 \\ 1 & 3 \end{pmatrix}^{-1} = \frac{1}{2} \begin{pmatrix} 3 & -1 \\ -1 & 1 \end{pmatrix}; \quad [\tilde{\Sigma}_{1,3}^{-1}]^V = \frac{1}{2} \begin{pmatrix} 3 & 0 & -1 \\ 0 & 0 & 0 \\ -1 & 0 & 1 \end{pmatrix} \\ \tilde{\Sigma}_{2,3}^{-1} &= \begin{pmatrix} 2 & 2 \\ 2 & 3 \end{pmatrix}^{-1} = \frac{1}{2} \begin{pmatrix} 3 & -2 \\ -2 & 2 \end{pmatrix}; \quad [\tilde{\Sigma}_{2,3}^{-1}]^V = \frac{1}{2} \begin{pmatrix} 0 & 0 & 0 \\ 0 & 3 & -2 \\ 0 & -2 & 2 \end{pmatrix} \\ [\tilde{\Sigma}_{3}^{-1}]^V &= \begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & \frac{1}{3} \end{pmatrix} \\ \tilde{K} &= \begin{pmatrix} \frac{3}{2} & 0 & -\frac{1}{2} \\ 0 & \frac{3}{2} & -1 \\ -\frac{1}{2} & -1 & \frac{7}{6} \end{pmatrix}; \quad \tilde{\Sigma} = \tilde{K}^{-1} = \begin{pmatrix} 1 & \frac{2}{3} & 1 \\ \frac{2}{3} & 2 & 2 \\ 1 & 2 & 3 \end{pmatrix} \end{split}$$

Exercise. Suppose that \mathcal{G} : 1—2—3, the mean $\xi = 0$ and $\tilde{\Sigma} = \begin{pmatrix} 1 & 1 & 0.9 \\ 1 & 2 & 2 \\ 0.9 & 2 & 3 \end{pmatrix}$. Compute by the clique-separator formula the MLEs \hat{K} and $\hat{\Sigma}$.

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Basic factorizations Maximum likelihood estimates An example

Mathematics marks



This graph is chordal with cliques $\{1, 2, 3\}$, $\{3, 4, 5\}$ with separator $S = \{3\}$ having $\nu(\{3\}) = 1$.

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Basic factorizations Maximum likelihood estimates An example

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Since one degree of freedom is lost by subtracting the average, we get in this example

$$\hat{K} = 87 \begin{pmatrix} w_{[123]}^{11} & w_{[123]}^{12} & w_{[123]}^{13} & 0 & 0 \\ w_{[123]}^{21} & w_{[123]}^{22} & w_{[123]}^{23} & 0 & 0 \\ w_{[123]}^{31} & w_{[123]}^{32} & w_{[123]}^{33} + w_{[345]}^{33} - 1/w_{33} & w_{[345]}^{34} & w_{[345]}^{35} \\ 0 & 0 & w_{[345]}^{43} & w_{[345]}^{44} & w_{[345]}^{45} \\ 0 & 0 & w_{[345]}^{53} & w_{[345]}^{54} & w_{[345]}^{55} \end{pmatrix}$$

where $w_{[123]}^{ij}$ is the *ij*th element of the inverse of

$$W_{[123]} = \begin{pmatrix} w_{11} & w_{12} & w_{13} \\ w_{21} & w_{22} & w_{23} \\ w_{31} & w_{32} & w_{33} \end{pmatrix}$$

and so on.