

# Decomposition of log-linear models

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A density  $f$  *factorizes* w.r.t.  $\mathcal{A}$  if there exist functions  $\psi_a(x)$  which depend on  $x_a$  only so that

$$f(x) = \prod_{a \in \mathcal{A}} \psi_a(x).$$

The set of distributions  $\mathcal{P}_{\mathcal{A}}$  which factorize w.r.t.  $\mathcal{A}$  is the *hierarchical log-linear model* generated by  $\mathcal{A}$ .

$\mathcal{A}$  is the *generating class* of the log-linear model.

For any generating class  $\mathcal{A}$  we construct the dependence graph  $G(\mathcal{A}) = G(\mathcal{P}_{\mathcal{A}})$  of the log-linear model  $\mathcal{P}_{\mathcal{A}}$ .

*The dependence graph is determined by the relation*

$$\alpha \sim \beta \iff \exists a \in \mathcal{A} : \alpha, \beta \in a.$$

For sets in  $\mathcal{A}$  are clearly complete in  $G(\mathcal{A})$  and therefore *distributions in  $\mathcal{P}_{\mathcal{A}}$  do factorize according to  $G(\mathcal{A})$ .*

They are thus also global, local, and pairwise Markov w.r.t.  $G(\mathcal{A})$ .

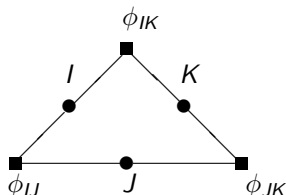
As a generating class defines a dependence graph  $G(\mathcal{A})$ , the reverse is also true.

The set  $\mathcal{C}(\mathcal{G})$  of *cliques* (maximal complete subsets) of  $\mathcal{G}$  is a generating class for the log-linear model of distributions which factorize w.r.t.  $\mathcal{G}$ .

If the dependence graph completely summarizes the restrictions imposed by  $\mathcal{A}$ , i.e. if

$$\mathcal{A} = \mathcal{C}(G(\mathcal{A})),$$

$\mathcal{A}$  is *conformal*.



The *factor graph* of  $\mathcal{A}$  is the bipartite graph with vertices  $V \cup \mathcal{A}$  and edges define by

$$\alpha \sim a \iff \alpha \in a.$$

Using this graph even non-conformal log-linear models admit a simple visual representation.

The maximum likelihood estimate  $\hat{p}$  of  $p$  is the unique element of  $\overline{\mathcal{P}_{\mathcal{A}}}$  which satisfies the system of equations

$$n\hat{p}(x_a) = n(x_a), \forall a \in \mathcal{A}, x_a \in \mathcal{X}_a. \quad (1)$$

Here  $g(x_a) = \sum_{y: y_a = x_a} g(y)$  is the *a-marginal* of the function  $g$ .  
The system of equations (1) expresses the *fitting of the marginals* in  $\mathcal{A}$ .

There is a *convergent* algorithm which solves the likelihood equations. This cycles (repeatedly) through all the  $a$ -marginals in  $\mathcal{A}$  and fit them one by one.

For  $a \in \mathcal{A}$  define the following *scaling* operation on  $p$ :

$$(T_a p)(x) \leftarrow p(x) \frac{n(x_a)}{np(x_a)}, \quad x \in \mathcal{X}$$

where  $0/0 = 0$  and  $b/0$  is undefined if  $b \neq 0$ .

Make an ordering of the generators  $\mathcal{A} = \{a_1, \dots, a_k\}$ . Define  $S$  by a full cycle of scalings

$$Sp = T_{a_k} \cdots T_{a_2} T_{a_1}.$$

Define the iteration

$$p_0(x) \leftarrow 1/|\mathcal{X}|, \quad p_n = Sp_{n-1}, n = 1, \dots$$

*It then holds that*

$$\lim_{n \rightarrow \infty} p_n = \hat{p}$$

*where  $\hat{p}$  is the unique maximum likelihood estimate of  $p \in \overline{\mathcal{P}_{\mathcal{A}}}$ , i.e. the solution of the equation system (1).*



*In some cases the IPS algorithm converges after a finite number of cycles.* An explicit formula is then available for the MLE of  $p \in \mathcal{P}_{\mathcal{A}}$ .

Consider first the case of a generating class with only two elements:  $\mathcal{A} = \{a, b\}$  and thus  $V = a \cup b$ . Let  $c = a \cap b$ . Recall that the MLE is the unique solution to

$$n\hat{p}(x_a) = n(x_a), \forall a \in \mathcal{A}, x_a \in \mathcal{X}_a.$$

Let

$$p^*(x) = \frac{n(x_a)n(x_b)}{n(x_c)n}.$$

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This satisfies (1) since e.g.

$$\begin{aligned} np^*(x_a) &= \sum_{y:y_a=x_a} \frac{n(y_a)n(y_b)}{n(y_c)} = \sum_{y:y_a=x_a} \frac{n(x_a)n(y_b)}{n(x_c)} \\ &= \frac{n(x_a)}{n(x_c)} \sum_{y:y_a=x_a} n(y_b) = \frac{n(x_a)}{n(x_c)} n(x_c) = n(x_a) \end{aligned}$$

and similarly with the other marginal. Hence we have  $\hat{p} = p^*$ .

The generating class  $\mathcal{A} = \{a, b\}$  is conformal. Its dependence graph  $\mathcal{G}$  has exactly two cliques  $a$  and  $b$ .

The graph is *chordal*, meaning that any cycle of length  $\geq 4$  has a chord.

$\mathcal{A}$  is called *decomposable* if  $\mathcal{A}$  is conformal, i.e.  $\mathcal{A} = \mathcal{C}(\mathcal{G})$ , and  $\mathcal{G}$  is chordal.

*The IPS-algorithm converges after a finite number of cycles (at most two) if and only if  $\mathcal{A}$  is decomposable.*

## Non-decomposable generating classes

A generating class can be non-decomposable in different ways.

The generating class  $\mathcal{A} = \{\{1, 2\}, \{2, 3\}, \{1, 3\}\}$  is the smallest non-decomposable generating class. This is non-conformal.

The graph below is the smallest non-chordal graph and its generating class is non-decomposable:



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

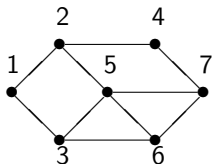
$$A \perp_{\mathcal{G}} B \mid S \text{ and } S \text{ is complete.}$$

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  $\mathcal{G}_{B \cup S}$ .

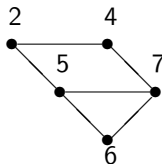
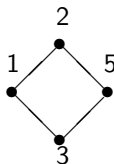
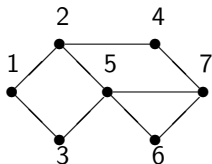
A graph is *prime* if no proper decomposition exists.

# Examples



The graph to the left is prime

Decomposition with  $A = \{1, 3\}$ ,  $B = \{4, 6, 7\}$  and  $S = \{2, 5\}$



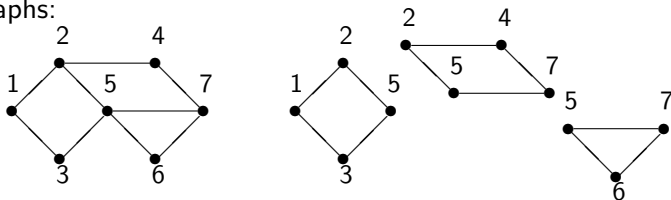
Suppose  $P$  satisfies (F) w.r.t.  $\mathcal{G}$  and  $(A, B, S)$  is a decomposition.  
Then

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

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.  $\mathcal{G}$ .*

# Decomposability

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



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*.



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  $\mathcal{S}$  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.

As we have a particularly simple factorization of the density, we have a similar factorization of the maximum likelihood estimate for a decomposable log-linear model.

*The MLE for  $p$  under the log-linear model with generating class  $\mathcal{A} = \mathcal{C}(\mathcal{G})$  for a chordal graph  $\mathcal{G}$  is*

$$\hat{p}(x) = \frac{\prod_{C \in \mathcal{C}} n(x_C)}{n \prod_{S \in \mathcal{S}} n(x_S)^{\nu(S)}}$$

*where  $\nu(S)$  is the number of times  $S$  appears as a separator in the total decomposition of its dependence graph.*

## Perfect numbering

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

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

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

## 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  $(\alpha, \beta)$ -separator are complete.

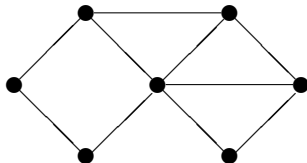
*Trees are chordal graphs* and thus decomposable.

Here is a (greedy) algorithm for checking chordality:

1. Look for a vertex  $v^*$  with  $\text{bd}(v^*)$  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.*

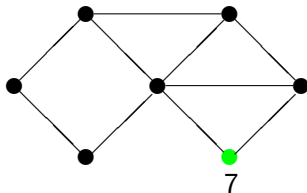
The complexity of this algorithm is  $O(|V|^2)$ .

# Greedy algorithm



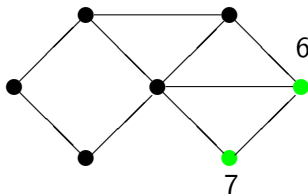
Is this graph chordal?

# Greedy algorithm



Is this graph chordal?

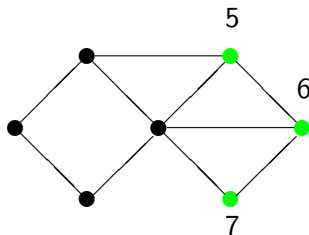
# Greedy algorithm



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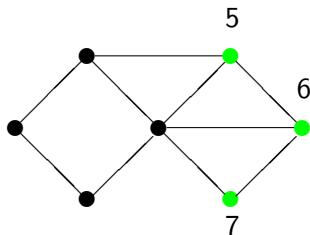


## Greedy algorithm



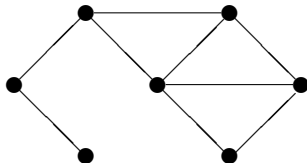
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## Greedy algorithm



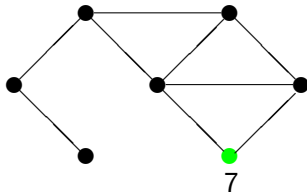
This graph is *not* chordal, as there is no candidate for number 4.

# Greedy algorithm



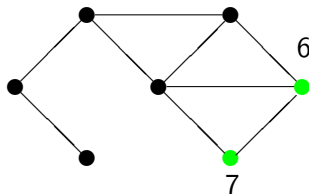
Is this graph chordal?

# Greedy algorithm



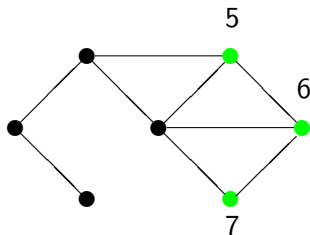
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# Greedy algorithm



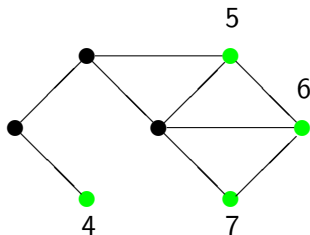
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## Greedy algorithm



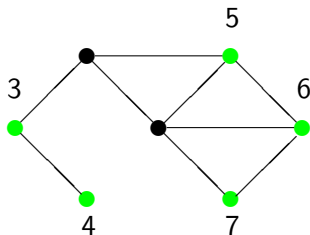
Is this graph chordal?

## Greedy algorithm



Is this graph chordal?

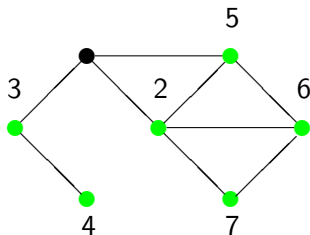
## Greedy algorithm



Is this graph chordal?

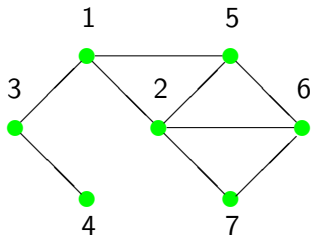


## Greedy algorithm



Is this graph chordal?

## Greedy algorithm

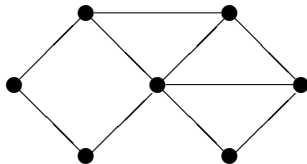


*This graph is chordal!*

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

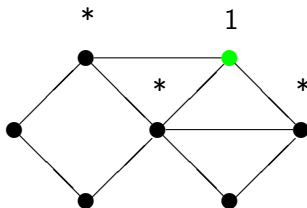
1. Choose  $v_0 \in V$  arbitrary and let  $v_0 = 1$ ;
2. When vertices  $\{1, 2, \dots, j\}$  have been identified, choose  $v = j + 1$  among  $V \setminus \{1, 2, \dots, j\}$  with highest cardinality of its numbered neighbours;
3. *If  $bd(j + 1) \cap \{1, 2, \dots, j\}$  is not complete,  $\mathcal{G}$  is not chordal;*
4. Repeat from 2;
5. *If the algorithm continues until only one vertex is left, the graph is chordal and the numbering is perfect.*

# Maximum Cardinality Search



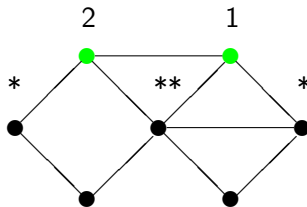
Is this graph chordal?

# Maximum Cardinality Search



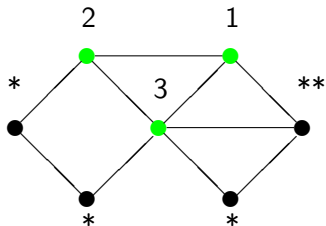
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# Maximum Cardinality Search



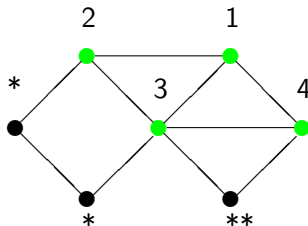
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# Maximum Cardinality Search



Is this graph chordal?

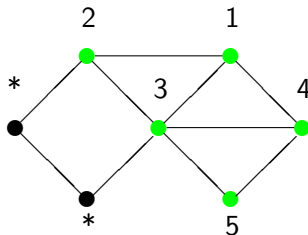
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Is this graph chordal?

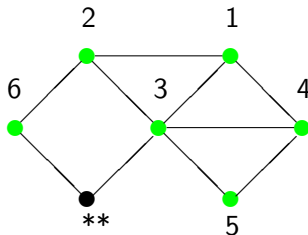


# Maximum Cardinality Search



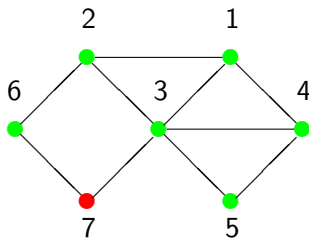
Is this graph chordal?

# Maximum Cardinality Search



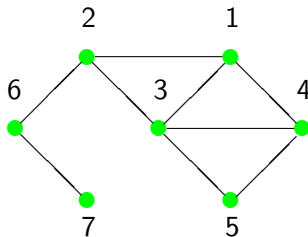
Is this graph chordal?

# Maximum Cardinality Search



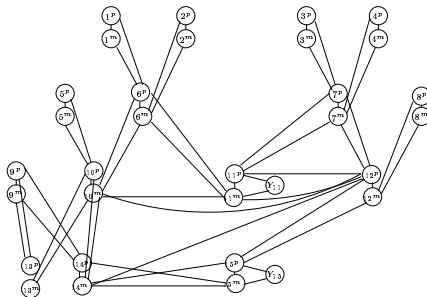
*The graph is not chordal!* because 7 does not have a complete boundary.

# Maximum Cardinality Search



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

# A chordal graph



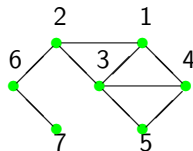
This graph is chordal, but it might not be that easy to see. . . Maximum Cardinality Search is handy!

## Finding the cliques of a chordal graph

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

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

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



$\pi_\lambda: 0, 1, 2, 2, 2, 1, 1.$

The cliques are  $C_\lambda = \{\lambda\} \cup B_\lambda, \lambda \in \Lambda.$