

week 11 day 1

Graphical models: continuation
Algebraic & Geometric Methods in Statistics

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Created for Math/Stat 561

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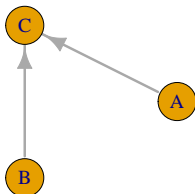
- Chapter 13: Graphical models
- We are following after Miles Bakenhus' course project lecture on sections 13.1 and 13.2.
- We will review a couple of examples from the basics of graphical models (think of the first st of these slides as your study worksheet in class).
- We will then see a few more examples
- Discuss the *discrete distributions* and connection to algebra&geometry.

Examples

- Genes:
 - three genes in this example A,B,C
- Relationships:
 - A regulates C
 - B regulates C

BIOLOGY

- genes
- relationships



GRAPH

- vertices
- edges

$$P(A, B, C) = P(A)P(B)P(C|A, B)$$

PROBABILISTIC MODEL

- random variables
- statistical dependencies

Correlation vs causation

- Genes regulated as $X \rightarrow Y \rightarrow Z$
- Z and X are correlated, but do not interact directly

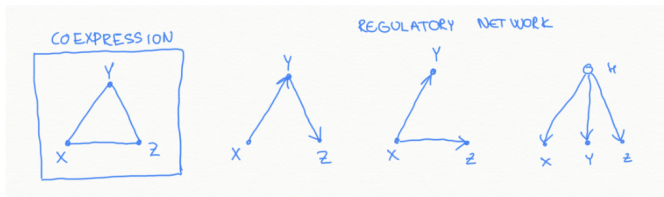


Figure 1: Source: K. Kubjas

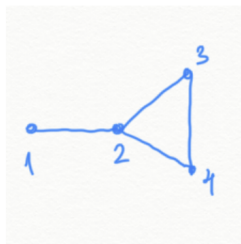
Separator

Poll:

Let G be a graph with nodes $\{1,2,3,4\}$ and edges $(1,2)$, $(2,3)$, $(2,4)$, $(3,4)$.

Which of the following sets are separators for the nodes 1 and 4?

- 1 $\{2\}$
- 2 $\{3\}$
- 3 $\{2,3\}$
- 4 $\{1,2,3,4\}$



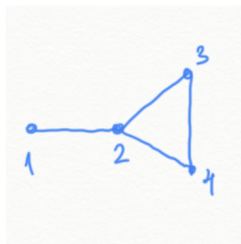
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Answer

Correct answers: 1. and 3.

Reminder: conditional independence definition

[board]

Pairwise Markov property

Let $G = (V, E)$ be an undirected graph.

Definition

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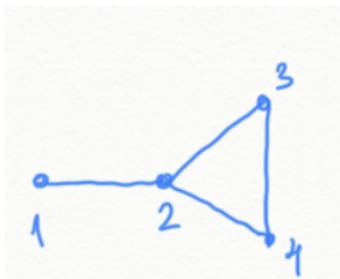
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Question (example)

The pairwise Markov property associated to G below is:

- ① $\{1 \perp\!\!\!\perp 3 | (2, 4), 1 \perp\!\!\!\perp 4 | (2, 3)\}$
- ② $\{1 \perp\!\!\!\perp 3 | 2, 1 \perp\!\!\!\perp 4 | 2\}$
- ③ $\{1 \perp\!\!\!\perp 3 | (2, 4)\}$
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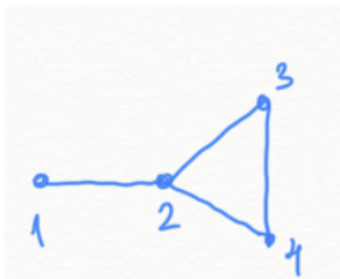
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Correct answer: 1.



Multivariate Gaussian random variables

- The CI statement $X_u \perp\!\!\!\perp X_v | X_{V \setminus \{u,v\}}$ is equivalent to the matrix $\Sigma_{V \setminus \{u\}, V \setminus \{v\}}$ having rank $|V \setminus \{u, v\}|$, or equivalently $\det(\Sigma_{V \setminus \{u\}, V \setminus \{v\}}) = 0$.
- This is equivalent to $(\Sigma^{-1})_{u,v} = 0$.
- The pairwise Markov property holds for a Gaussian distribution if and only if the entries of the concentration matrix corresponding to non-edges are zero.

Question (example)

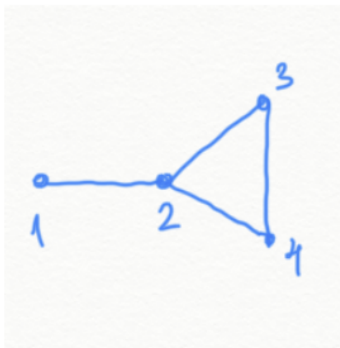
What is the form of the concentration matrices of a Gaussian distribution obeying the pairwise Markov property have?

1

$$\begin{bmatrix} k_{11} & 0 & k_{13} & k_{14} \\ 0 & k_{22} & 0 & 0 \\ k_{13} & 0 & k_{33} & 0 \\ k_{14} & 0 & 0 & k_{44} \end{bmatrix}$$

2

$$\begin{bmatrix} k_{12} & k_{12} & 0 & 0 \\ k_{12} & k_{22} & k_{23} & k_{24} \\ 0 & k_{23} & k_{33} & k_{34} \\ 0 & k_{24} & k_{34} & k_{44} \end{bmatrix}$$



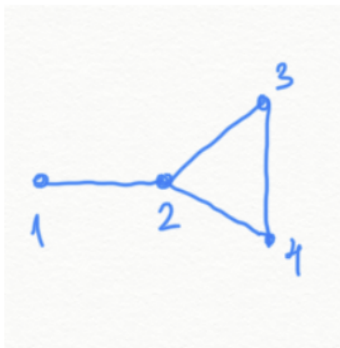
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Correct answer: 2.



Global Markov property

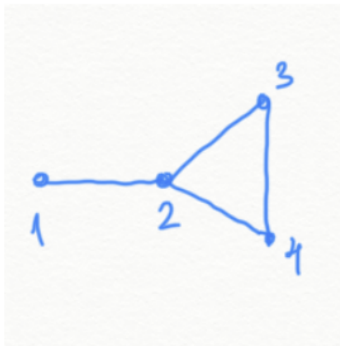
Definition (reminder!)

[board]

Question (example)

The global Markov property associated to G is:

- ① $\{1 \perp\!\!\!\perp (3, 4) | 2\}$
- ② $\{1 \perp\!\!\!\perp 3 | (2, 4), \quad 1 \perp\!\!\!\perp 4 | (2, 3)\}$
- ③ $\{1 \perp\!\!\!\perp 3 | (2, 4), \quad 1 \perp\!\!\!\perp 4 | (2, 3),$
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Global Markov property

Definition (reminder!)

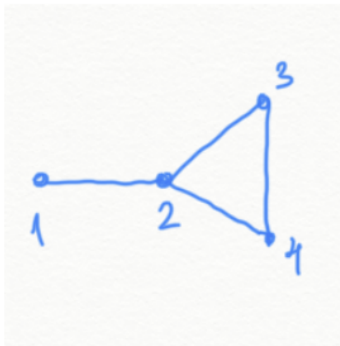
[board]

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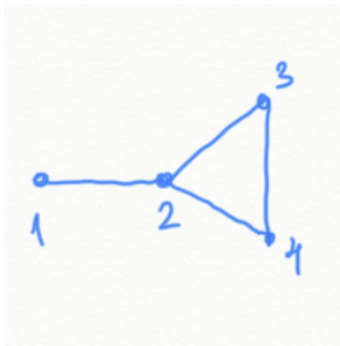
Correct answer: 3.



Markov properties

In the last lecture, Miles showed that pairwise Markov statements C_{pairs} are a subset of Global statements C_{global} .

- In our example:
- $C_{pairs} = \{1 \perp\!\!\!\perp 3|(2, 4), \quad 1 \perp\!\!\!\perp 4|(2, 3)\}.$
- $C_{global} = C_{pairs} \cup \{1 \perp\!\!\!\perp (3, 4)|2\}.$



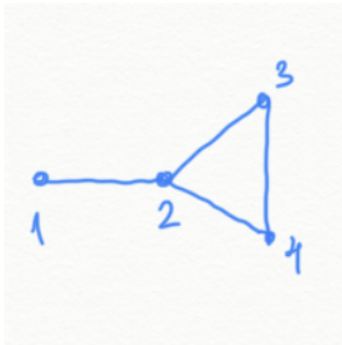
Factorization property

- We want to characterize **all** the distributions that satisfy the Markov properties for a *given graph*.
 - Hammersley-Clifford theorem relates the implicit description of a graphical model through Markov properties to a parametric description.
- Recall: definition of factorizing according to a graph via cliques.
[board]
- Review Theorem 13.2.10 (recursive factorization in DAGs) with *proof*.

Question (example)

What are the **maximal cliques** of G ?

- ① $\{1\}$
- ② $\{1, 2\}$
- ③ $\{1, 2, 3, \}$
- ④ $\{2, 3, 4\}$

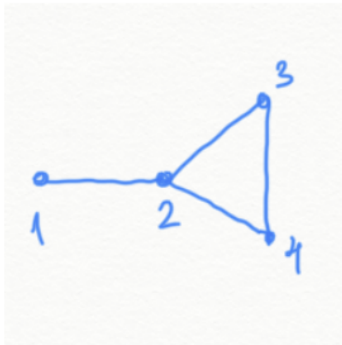


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- ③ $\{1, 2, 3, \}$
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Correct answers: 2 and 4.



Examples from 13.4

- (Homogeneous) Markov chain - example 13.4.1 and connection to chapter 1
- Hidden Markov model

[board notes]

Discrete distributions - and connection to algebra and geometry

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This document is created for Math/Stat 561, Spring 2023.

The slides that are not directly from the book are sourced from **Kaie Kubjas'** Algebraic Statistics course at Aalto University.

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