Tobias Boege

The Gaussian CI inference problem

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Consider random variables $(\xi_i)_{i \in N}$. The conditional independence (CI) statement $\xi_i \perp \xi_j \mid \xi_K$ conveys, informally, that if $\xi_K$ is known, then learning the value of $\xi_i$ does not give any information about $\xi_j$. 

Example: Let $c_1$ and $c_2$ be two independent coins and $b$ a bell which rings if and only if $c_1$ and $c_2$ land with the same side up. What is the conditional independence relation of the system $(c_1, c_2, b)$ of random variables? 

$\rightarrow c_1 \perp c_2$ and $\neg (c_1 \perp c_2 \mid b)$ and ...
Consider random variables $(\xi_i)_{i \in N}$. The conditional independence (CI) statement $\xi_i \independent \xi_j \mid \xi_K$ conveys, informally, that if $\xi_K$ is known, then learning the value of $\xi_i$ does not give any information about $\xi_j$.

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Consider random variables \((\xi_i)_{i \in \mathbb{N}}\). The conditional independence (CI) statement \(\xi_i \perp \xi_j \mid \xi_K\) conveys, informally, that if \(\xi_K\) is known, then learning the value of \(\xi_i\) does not give any information about \(\xi_j\).

**Example:** Let \(c_1\) and \(c_2\) be two independent coins and \(b\) a bell which rings if and only if \(c_1\) and \(c_2\) land with the same side up. What is the conditional independence relation of the system \((c_1, c_2, b)\) of random variables? \(\rightarrow c_1 \perp c_2\) and \(\neg(c_1 \perp c_2 \mid b)\) and \ldots
Gaussian conditional independence

Let the random vector be normally distributed: $(\xi_i)_{i \in \mathbb{N}} \sim \mathcal{N}(\mu, \Sigma)$.

**Definition**

The polynomial $\Sigma[K] := \text{det} \Sigma_{K,K}$ is a principal minor of $\Sigma$ and $\Sigma[ij|K] := \text{det} \Sigma_{iK,jK}$ is an almost-principal minor.

If $\Sigma$ is positive-definite, then $\Sigma[K] > 0$, and $\xi_i \perp \xi_j \mid \xi_K$ holds if and only if $\Sigma[ij|K] = 0$. 
almost-principal minors

\[ \Sigma[ij] = x_{ij} \]
\[ \Sigma[ij|k] = x_{ij}x_{kk} - x_{ik}x_{jk} \]
\[ \Sigma[ij|kl] = x_{ij}x_{kk}x_{ll} - x_{il}x_{jl}x_{kk} + x_{il}x_{jk}x_{kl} + x_{ik}x_{jl}x_{kl} - x_{ij}x_{kl}^2 - x_{ik}x_{jk}x_{ll} \]
\[ \Sigma[ij|klm] = x_{ij}x_{kk}x_{ll}x_{mm} + x_{im}x_{jm}x_{kl}^2 - x_{im}x_{jl}x_{kl}x_{km} - x_{il}x_{jm}x_{kl}x_{km} + x_{il}x_{jl}x_{km}^2 \]
\[ - x_{im}x_{jm}x_{kk}x_{ll} + x_{im}x_{jk}x_{km}x_{ll} + x_{ik}x_{jm}x_{km}x_{ll} - x_{ij}x_{km}^2x_{ll} \]
\[ + x_{im}x_{jl}x_{kk}x_{lm} + x_{il}x_{jm}x_{kk}x_{lm} - x_{im}x_{jk}x_{kl}x_{lm} - x_{ik}x_{jm}x_{kl}x_{lm} \]
\[ - x_{il}x_{jk}x_{km}x_{lm} - x_{ik}x_{jl}x_{km}x_{lm} + 2x_{ij}x_{kl}x_{km}x_{lm} + x_{ik}x_{jk}x_{lm}^2 \]
\[ - x_{ij}x_{kk}^2x_{lm} - x_{il}x_{jl}x_{kk}x_{mm} + x_{il}x_{jk}x_{kl}x_{mm} + x_{ik}x_{jl}x_{kl}x_{mm} \]
\[ - x_{ij}x_{kl}^2x_{mm} - x_{ik}x_{jk}x_{ll}x_{mm} \]
\[ \vdots \]
Gaussian CI models

Definition

A CI constraint is a CI statement $\xi_i \perp \xi_j \mid \xi_K$ or its negation $\neg(\xi_i \perp \xi_j \mid \xi_K)$. They are algebraic conditions on the entries of $\Sigma$, equivalent to vanishing or non-vanishing of the almost-principal minors $\Sigma[ij|K]$.

Definition

The model of a set of CI constraints is the set of all positive-definite matrices which satisfy the constraints.
Figure: Model of $\Sigma[12|3] = 0$ in the space of $3 \times 3$ correlation matrices.
Consider two sets of CI statements $\mathcal{P}$ and $\mathcal{Q}$:

$$\bigwedge \mathcal{P} \Rightarrow \bigvee \mathcal{Q}$$
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\[
\bigwedge \mathcal{P} \Rightarrow \bigvee \mathcal{Q}
\]

is not valid

\[
\mathcal{P} \cup \neg \mathcal{Q}
\]

is non-empty

Reasoning about relevance statements in normally distributed random variables is the same as reasoning about the vanishing of very special kinds of determinants on very special kinds of varieties inside the positive-definite matrices.
Consider two sets of CI statements $\mathcal{P}$ and $\mathcal{Q}$:

$$\wedge \mathcal{P} \Rightarrow \vee \mathcal{Q}$$

is not valid

$$\iff$$

$$\mathcal{P} \cup \neg \mathcal{Q}$$

is non-empty

Reasoning about relevance statements in normally distributed random variables is the same as reasoning about the vanishing of very special kinds of determinants on very special kinds of varieties inside the positive-definite matrices.
For geometers: conditional independence $\approx$ collinearity
Examples of CI inference

Consider a general positive-definite $3 \times 3$ correlation matrix

$$\Sigma = \begin{pmatrix} 1 & a & b \\ a & 1 & c \\ b & c & 1 \end{pmatrix}.$$ 

- If $\Sigma[12|3] = a - bc$ and $\Sigma[13|1] = b$ vanish, then $\Sigma[12|1] = a$ and $\Sigma[13|2] = b - ac$ must vanish as well:
  $$(12|3) \wedge (13|1) \Rightarrow (12|1) \wedge (13|2).$$
Examples of CI inference

\[ \Sigma = \begin{pmatrix} 1 & a & b \\ a & 1 & c \\ b & c & 1 \end{pmatrix} \]

- If \( \Sigma_{12} = a \) and \( \Sigma_{123} = a - bc \) vanish, then \( bc = \Sigma_{13} \cdot \Sigma_{23} \) must vanish:

\[
(12) \land (123) \Rightarrow (13) \lor (23).
\]
Rational points on CI models

Šimeček’s Question (2006)

Does every non-empty Gaussian CI model contain a rational point?

Or: can every wrong inference rule be refuted over $\mathbb{Q}$?
Rational points on CI models

Šimeček’s Question (2006)

*Does every non-empty Gaussian CI model contain a rational point?*

Or: can every wrong inference rule be refuted over \( \mathbb{Q} \)?

**Model M85**

\[
\begin{aligned}
    a &= \frac{3}{632836} \sqrt{1107463}, \\
    b &= 10c = \frac{100}{158209} \sqrt{1107463}, \\
    d &= 10e = \frac{3}{4}, f = \frac{1}{10}
\end{aligned}
\]

\[
\begin{pmatrix}
    1 & -1/17 & -49/51 & -7/17 \\
    -1/17 & 1 & 1/3 & 1/7 \\
    -49/51 & 1/3 & 1 & 3/7 \\
    -7/17 & 1/7 & 3/7 & 1
\end{pmatrix}
\]
Complexity bounds from real geometry

Let $f_i \in \mathbb{Z}[t_1, \ldots, t_k]$ be integer polynomials in finitely many variables.

**Theorem (Tarski’s transfer principle)**

*If a polynomial system $\{f_i \# i; 0\}$, where $\# i \in \{=, \neq, <, \leq, \geq, >\}$, has a solution over $\mathbb{R}$, then it has a solution in a finite real extension of $\mathbb{Q}$."

$\rightarrow$ If $\wedge P \Rightarrow \vee Q$ is false, there exists a counterexample matrix $\Sigma$ with algebraic entries.

$(12|) \wedge (12|3) \Rightarrow (13|)$ is false and a counterexample is

$$
\begin{pmatrix}
1 & 0 & 1/2 \\
0 & 1 & 0 \\
1/2 & 0 & 1
\end{pmatrix}.
$$
Complexity bounds from real geometry

Let \( F, f_i, g_j \in \mathbb{Z}[t_1, \ldots, t_k] \) be integer polynomials in finitely many variables.

**Theorem (Positivstellensatz)**

A polynomial \( F \) vanishes on the basic semialgebraic set \( \{ f_i = 0, g_j \geq 0 \} \) if and only if 
\[-F^{2m} \in \text{ideal}(f_i) + \text{cone}(g_j) \text{ for } m \text{ large enough}.\]

→ If \( \land P \Rightarrow \lor Q \) is true, there exists an algebraic proof for it with rational coefficients.

\((12|) \land (12|3) \Rightarrow (13|) \lor (23|)\) is true and a proof is the polynomial identity
\[
\Sigma[13|] \cdot \Sigma[23|] = \Sigma[3] \cdot \Sigma[12|] - \Sigma[12|3].
\]

The associated decision problem is the *existential theory of the reals*. 
Universality theorems

Theorem (B. 2021)

For every finite real extension $K/Q$ there exists a Gaussian CI model $M_K$ such that:
for every $L/Q$, $M_K$ has an $L$-rational point if and only if $K \subseteq L$.

→ The answer to Šimeček’s question is NO.

Theorem (B. 2021)

The problem of deciding whether a CI inference formula is valid for all Gaussian distributions
is polynomial-time equivalent to the existential theory of the reals.
Theorem

To every polynomial system \( \{ f_i \neq 0 \} \) one can compute a polynomially-sized set of CI constraints which has a model over a real algebraic extension \( K/Q \) if and only if the polynomial system has a solution in \( K \).
Point and line configuration for the equation 
\[ x^2 - 2 = 0. \]

The configuration is specified by incidences between 
points and lines and also the parallelities of lines.

It is realizable over \( \mathbb{Q}(\sqrt{2}) \) but not over \( \mathbb{Q} \).

Keyword for the general technique: von Staudt constructions (1857).
\( \Sigma[ij] = x_{ij} \rightarrow \text{impose } x_{kl} = x_{km} = x_{lm} = 0 \) on a correlation matrix, then:

\[
\Sigma[ij|klm] = x_{ij} x_{kk} x_{ll} x_{mm} + x_{im} x_{jm} x_{kl}^2 - x_{im} x_{jl} x_{kl} x_{km} - x_{il} x_{jm} x_{kl} x_{km} + x_{il} x_{jl} x_{kl}^2 \\
- x_{im} x_{jm} x_{kk} x_{ll} + x_{im} x_{jk} x_{km} x_{ll} + x_{ik} x_{jm} x_{km} x_{ll} - x_{ij} x_{km}^2 x_{ll} \\
+ x_{im} x_{jl} x_{kk} x_{lm} + x_{il} x_{jm} x_{kk} x_{lm} - x_{im} x_{jk} x_{kl} x_{lm} - x_{ik} x_{jm} x_{kl} x_{lm} \\
- x_{il} x_{jk} x_{km} x_{lm} - x_{ik} x_{jl} x_{km} x_{lm} + 2 x_{ij} x_{kl} x_{km} x_{lm} + x_{ik} x_{jl} x_{kl}^2 \\
- x_{ij} x_{kk} x_{lm}^2 - x_{il} x_{jl} x_{kk} x_{mm} + x_{il} x_{jk} x_{kl} x_{mm} + x_{ik} x_{jl} x_{kl} x_{mm} \\
- x_{ij} x_{kl}^2 x_{mm} - x_{ik} x_{jk} x_{ll} x_{mm}
\]

\[
= x_{ij} - \sum_{t=k,l,m} x_{it} x_{jt} = x_{ij} - \left( \begin{pmatrix} x_{ik} \\ x_{il} \\ x_{im} \end{pmatrix}, \begin{pmatrix} x_{jk} \\ x_{jl} \\ x_{jm} \end{pmatrix} \right). 
\]
Incidence relation in a CI model

\[
\begin{pmatrix}
  p_1 & \ldots & p_n & l_1 & \ldots & l_m & x & y & z \\
p_1^* & \langle p, p' \rangle & l_1^* & \langle \ell, \ell' \rangle & p_1^x & p_1^y & p_1^z \\
\vdots & & \ddots & & \vdots & & \vdots & & \vdots \\
p_n & \langle p', p \rangle & p_n^* & \langle \ell, \ell \rangle & p_n^x & p_n^y & p_n^z \\
l_1 & \langle \ell, p \rangle & \ell_1^* & \ell_1^{x} & p_1^x & \ell_1^y & \ell_1^z \\
\vdots & & \ddots & & \vdots & & \vdots & & \vdots \\
l_m & \langle \ell', \ell \rangle & \ell_m^{x} & \ell_m^{y} & \ell_m^{z} & \ell_1^x & \ell_1^y & \ell_1^z \\
x & p_1^x & p_n^x & \ell_1^x & \ell_m^x & 1 & 0 & 0 \\
y & p_1^y & \ldots & p_n^y & \ell_1^y & \ell_m^y & 0 & 1 & 0 \\
z & p_1^z & p_n^z & \ell_1^z & \ell_m^z & 0 & 0 & 1
\end{pmatrix}
\]
Approximations to the inference problem
Approximations to the inference problem

Theorem (Matúš 2005)

The following relations hold for every symmetric matrix $\Sigma$:

$$\Sigma[kL] \cdot \Sigma[ij|L] = \Sigma[L] \cdot \Sigma[ij|kL] + \Sigma[ik|L] \cdot \Sigma[jk|L].$$

These relations define essential geometric properties of symmetric matrices in principal and almost-principal minor coordinates. Study their combinatorics!
The Gaussian CI configuration space

\[ \Sigma[kL] \cdot \Sigma[ij|L] = \Sigma[L] \cdot \Sigma[ij|kL] + \Sigma[ik|L] \cdot \Sigma[jk|L] \]

The Gaussian CI configuration space \( G \subseteq \mathbb{R}^{2n} \times \mathbb{R}^{(n)^2_{2n-2}} \) consists of all vectors of principal and almost-principal minors of \( \Sigma \in \text{PD}_n \).
The Gaussian CI configuration space

\[ \Sigma[kL] \cdot \Sigma[ijL] = \Sigma[L] \cdot \Sigma[ij|kL] + \Sigma[ik|L] \cdot \Sigma[jk|L] \]

The Gaussian CI configuration space \( \mathcal{G} \subseteq \mathbb{R}^{2n} \times \mathbb{R}^{(n^2)}_{2n-2} \) consists of all vectors of principal and almost-principal minors of \( \Sigma \in \text{PD}_n \).

Very wasteful encoding of a matrix, but this creates simple and useful relations on configuration vectors. The CI structure of \( \Sigma \) is encoded in the zero pattern of \( c(\Sigma) \in \mathcal{G} \).
Combinatorial compatibility

\[ \Sigma[kL] \cdot \Sigma[ij|L] = \Sigma[L] \cdot \Sigma[ij|kL] + \Sigma[ik|L] \cdot \Sigma[jk|L] \]

Combinatorial compatibility means fulfilling polynomial relations under uncertainty:
What if we only knew that all \( \Sigma[K] \neq 0 \) and whether or not \( \Sigma[ij|K] = 0 \) for every \( (ij|K) \)?
Combinatorial compatibility

\[
\Sigma[kL] \cdot \Sigma[ij|L] = \Sigma[L] \cdot \Sigma[ij|kL] + \Sigma[ik|L] \cdot \Sigma[jk|L]
\]

*Combinatorial compatibility* means fulfilling polynomial relations under uncertainty:

What if we only knew that all \(\Sigma[K] \neq 0\) and whether or not \(\Sigma[ij|K] = 0\) for every \((ij|K)\)?

\[
(ij|L) \land (ij|kL) \Rightarrow (ik|L) \lor (jk|L)
\]

\[
(ik|L) \land (ij|kL) \Rightarrow (ij|L)
\]

\[\vdots\]
Combinatorial compatibility

\[ \Sigma[kL] \cdot \Sigma[ij|L] = \Sigma[L] \cdot \Sigma[ij|kL] + \Sigma[ik|L] \cdot \Sigma[jk|L] \]

Combinatorial compatibility means fulfilling polynomial relations under uncertainty:
What if we only knew that all \( \Sigma[K] \neq 0 \) and whether or not \( \Sigma[ij|K] = 0 \) for every \( (ij|K) \)?

\[(ij|L) \land (ij|kL) \Rightarrow (ik|L) \lor (jk|L)\]
\[(ik|L) \land (ij|kL) \Rightarrow (ij|L) \land (ik|jL)\]
\[(ij|kL) \land (ik|jL) \Rightarrow (ij|L) \land (ik|L)\]
\[(ij|L) \land (ik|L) \Rightarrow (ij|kL) \land (ik|jL)\]

This yields the definition of gaussoids.
CI inference via SAT solvers

Since gaussoids have a finite axiomatization, a SAT solver like CaDiCaL can deduce implications under the gaussoid axioms:

\[(12|3) \land (12|34) \land (24|1) \land (34|2) \Rightarrow (12|) \land (12|4) \land (24|) \land (24|3) \land (24|13) \land (34|)\]

These conclusions are valid for all regular Gaussian distributions.
Oriented gaussoids

\[ \Sigma[kL] \cdot \Sigma[ij|L] = \Sigma[L] \cdot \Sigma[ij|kL] + \Sigma[ik|L] \cdot \Sigma[jk|L] \]

What if we only knew that all \( \text{sgn} \Sigma[K] = +1 \) and the value of \( \text{sgn} \Sigma[ij|K] \) for every \((ij|K)\)?

\[ + (ij|L) \land -(ij|kL) \Rightarrow [+(ik|L) \land +(jk|L)] \lor [-(ik|L) \land -(jk|L)] \]

→ Oriented and orientable gaussoids.
Oriented gaussoids

$$\Sigma[kL] \cdot \Sigma[ij|L] = \Sigma[L] \cdot \Sigma[ij|kL] + \Sigma[ik|L] \cdot \Sigma[jk|L]$$

What if we only knew that all \(\text{sgn} \, \Sigma[K] = +1\) and the value of \(\text{sgn} \, \Sigma[ij|K]\) for every \((ij|K)\)?

\[+(ij|L) \land -(ij|kL) \Rightarrow [+(ik|L) \land +(jk|L)] \lor [-(ik|L) \land -(jk|L)]\]

\(\Rightarrow\) Oriented and orientable gaussoids.

\[(ij|L) \land (kl|L) \land (ik|jL) \land (jl|ikL) \Rightarrow (ik|L)\]

\[(ij|L) \land (kl|iL) \land (kl|jL) \land (ij|kL) \Rightarrow (kl|L)\]

\[(ij|L) \land (jl|kL) \land (kl|iL) \land (ik|jlL) \Rightarrow (ik|L)\]

\[(ij|kL) \land (ik|jL) \land (il|iL) \Rightarrow (ij|L)\]

\[(ij|kL) \land (ik|jL) \land (jl|iL) \land (kjl) \Rightarrow (ij|L)\]
Using the gaussoid axioms, we find:

\[(12|) \land (13|4) \land (14|5) \land (15|23) \land (23|5) \land (24|135) \land (34|12) \land (35|1) \land (45|2)\]

\[\Rightarrow\text{ nothing.}\]

The structure on the left is a gaussoid.
Running the SAT solver CaDiCaL on the definition of oriented gaussoids confirms that their supports satisfy

\[(12|) \land (13|4) \land (14|5) \land (15|23) \land (23|5) \land (24|135) \land (34|12) \land (35|1) \land (45|2)\]

\[\Rightarrow\] everything except \((25|K)\) for all \(K\).

The geometric model is that of a Markov network!
The search for inference rules

Inference rules help characterize the realizable CI structures:

- **3-variate**: 11 out of 64 by Matúš 2005.
- **4-variate**: 629 out of 16,777,216 by Lněnička and Matúš 2007.
- **5-variate**: open! (out of 1,208,925,819,614,629,174,706,176)
  - 254,826 gaussoids modulo symmetry
  - 87,834 of which are orientable gaussoids
  - 84,908 of which are selfadhesive orientable gaussoids.

Help wanted:
- Use information inequalities and linear programming.
- Tropical approximations and valuated gaussoids.
- Compute algebraic realization spaces.
- Find and certify real solutions to polynomial systems.
The search for inference rules

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Proof sketch
Proof sketch

Theorem

To every polynomial system \( \{ f_i \not\equiv 0 \} \) one can compute a polynomially-sized set of CI constraints which has a model over a real algebraic extension \( \mathbb{K}/\mathbb{Q} \) if and only if the polynomial system has a solution in \( \mathbb{K} \).
(1) **Algebra ⊆ Synthetic geometry**

Point and line configuration for the equation \(x^2 - 2 = 0\).

The configuration is specified by incidences between points and lines and also the parallelities of lines.

It is realizable over \(\mathbb{Q}(\sqrt{2})\) but not over \(\mathbb{Q}\).
Von Staudt constructions

Addition

Multiplication
Von Staudt constructions

Addition

Multiplication
Von Staudt constructions

Addition

Multiplication
Von Staudt constructions

Addition

Multiplication
Von Staudt constructions

Addition

Multiplication
Von Staudt constructions

Addition

Multiplication
Von Staudt constructions

Addition

Multiplication
The cube root of 4

\[ x^3 - 4 \]
(2) Synthetic geometry $\subseteq$ Gaussian CI

$$\Sigma[ij] = x_{ij} \rightarrow \text{impose } x_{kl} = x_{km} = x_{lm} = 0 \text{ on a correlation matrix, then:}$$

$$\Sigma[ij|klm] = x_{ij} x_{kk} x_{ll} x_{mm} + x_{im} x_{jm} x_{kl}^2 - x_{im} x_{jl} x_{km} - x_{il} x_{jm} x_{kl} x_{km} + x_{il} x_{jl} x_{km}^2$$

$$- x_{im} x_{jm} x_{kk} x_{ll} + x_{im} x_{jk} x_{km} x_{ll} + x_{ik} x_{jm} x_{km} x_{ll} - x_{ij} x_{km}^2 x_{ll}$$

$$+ x_{im} x_{jl} x_{kk} x_{lm} + x_{il} x_{jm} x_{kk} x_{lm} - x_{im} x_{jk} x_{kl} x_{lm} - x_{ik} x_{jm} x_{kl} x_{lm}$$

$$- x_{il} x_{jk} x_{km} x_{lm} - x_{ik} x_{jl} x_{km} x_{lm} + 2 x_{ij} x_{kl} x_{km} x_{lm} + x_{ik} x_{jl} x_{km}^2$$

$$- x_{ij} x_{kk} x_{lm}^2 - x_{il} x_{jl} x_{kk} x_{mm} + x_{il} x_{jk} x_{kl} x_{mm} + x_{ik} x_{jl} x_{kl} x_{mm}$$

$$- x_{ij} x_{kl} x_{mm}^2 - x_{ik} x_{jk} x_{ll} x_{mm}$$

$$= x_{ij} - \sum_{k=k,l,m} x_{ik} x_{jk} = x_{ij} - \left( \begin{pmatrix} x_{jk} \\ x_{il} \\ x_{jm} \end{pmatrix}, \begin{pmatrix} x_{jk} \\ x_{il} \\ x_{jm} \end{pmatrix} \right).$$
(2) Synthetic geometry ⊆ Gaussian CI

\[
\sum [ij|klm] = 0 \iff \Sigma_{ij} = (\Sigma_{i,klm}, \Sigma_{j,klm}) \\
\sum [ij] = 0 \iff \Sigma_{ij} = 0
\]

\[
\iff \Sigma_{i,klm} \perp \Sigma_{j,klm}
\]
(2) Synthetic geometry $\subseteq$ Gaussian CI

\[
\Sigma[ij|klm] = 0 \iff \Sigma_{ij} = \langle \Sigma_{i,klm}, \Sigma_{j,klm} \rangle \\
\Sigma[ij] = 0 \iff \Sigma_{ij} = 0
\]
\[
\iff \Sigma_{i,klm} \perp \Sigma_{j,klm}
\]

\[p = \Sigma_{i,klm} = [p_x : p_y : p_z] \text{ and } \ell = \Sigma_{j,klm} = [\ell_x : \ell_y : \ell_z]\]

are the homogeneous coordinates of a point and a line in the projective plane with $p \in \ell^\perp$.  


(2) Synthetic geometry ⊆ Gaussian CI

\[ \Sigma[ij|klm] = 0 \iff \Sigma_{ij} = \langle \Sigma_{i,klm}, \Sigma_{j,klm} \rangle \]
\[ \Sigma[ij] = 0 \iff \Sigma_{ij} = 0 \]

\( p = \Sigma_{i,klm} = [p_x : p_y : p_z] \) and \( \ell = \Sigma_{j,klm} = [\ell_x : \ell_y : \ell_z] \)
are the homogeneous coordinates of a point and a line in the projective plane with \( p \in \ell^\perp \).
Incidence relation in a CI model

\[
\begin{pmatrix}
p_1 & \cdots & p_n & l_1 & \cdots & l_m & x & y & z \\
p_1^* & \langle p, p' \rangle & l_1^* & \cdots & \langle p, \ell \rangle & p_1^x & p_1^y & p_1^z \\
\vdots & \ddots & \vdots & \ddots & \ddots & \vdots & \vdots & \vdots & \vdots \\
p_n & \langle p', p \rangle & p_n^* & \langle \ell, p \rangle & \langle \ell, \ell' \rangle & p_n^x & p_n^y & p_n^z \\
l_1 & \vdots & \langle \ell, p \rangle & \ell_1^* & \cdots & \ell_1^x & \ell_1^y & \ell_1^z \\
\vdots & \ddots & \ell_1 & \ddots & \cdots & \ell_1^x & \ell_1^y & \ell_1^z \\
l_m & \langle \ell', \ell \rangle & l_m^* & \ell_m^x & \ell_m^y & \ell_m^z & \ell_m^x & \ell_m^y & \ell_m^z \\
x & p_1^x & \cdots & p_n^x & \ell_1^x & \cdots & \ell_m^x & 1 & 0 & 0 \\
y & p_1^y & \cdots & p_n^y & \ell_1^y & \cdots & \ell_m^y & 0 & 1 & 0 \\
z & p_1^z & \cdots & p_n^z & \ell_1^z & \cdots & \ell_m^z & 0 & 0 & 1 \\
\end{pmatrix}
\]