

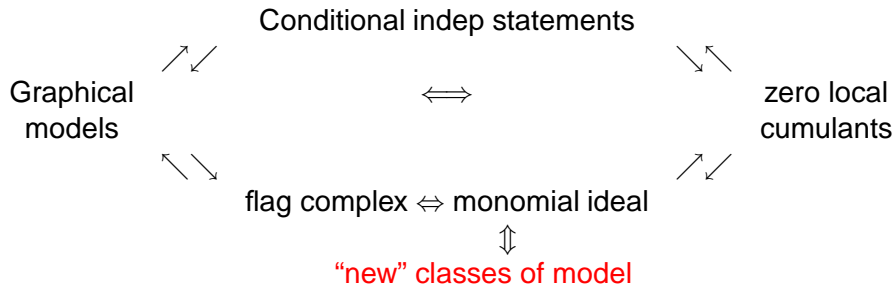
Hierarchical models and monomial ideals

Daniel Bruynooghe and Henry Wynn

London School of Economics
H.Wynn@lse.ac.uk

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Summary diagram



Local cumulants and monomial ideals

- Notation
- $\log f$ and local cumulants
- Independence and conditional independence
- Hierarchical models
- The duality with monomial ideals
- Some “new” examples

Notation

- Random variables: X_1, \dots, X_p
- pdf: $f(X) = f(X_1, \dots, X_n)$
- $g = \log f$
- Index set: $J \in N = \{1, \dots, p\}$
- Margin: X_J
- Multi-index: $\alpha = (\alpha_1, \dots, \alpha_p)$, $|\alpha| = \sum \alpha_i$
- Monomial: $x^\alpha = x_1^{\alpha_1} \dots x_p^{\alpha_p}$
- Differential:

$$D^\alpha = \frac{\partial^{|\alpha|}}{\partial x_1^{\alpha_1} \dots \partial x_p^{\alpha_p}}$$

Ordinary moments and cumulants

$$\kappa_{100} = \mu_{100}$$

$$\kappa_{010} = \mu_{010}$$

$$\kappa_{001} = \mu_{010}$$

$$\kappa_{110} = \mu_{110} - \mu_{100}\mu_{010}$$

$$\kappa_{101} = \mu_{101} - \mu_{010}\mu_{101}$$

$$\kappa_{011} = \mu_{011} - \mu_{010}\mu_{001}$$

$$\kappa_{111} = \mu_{111} - \mu_{100}\mu_{011} - \mu_{010}\mu_{101} - \mu_{110}\mu_{001} + 2\mu_{100}\mu_{010}\mu_{001}$$

$$\mu_{100} = \kappa_{100}$$

$$\mu_{010} = \kappa_{010}$$

$$\mu_{001} = \kappa_{010}$$

$$\mu_{110} = \kappa_{110} + \kappa_{100}\kappa_{010}$$

$$\mu_{101} = \kappa_{101} + \kappa_{010}\kappa_{101}$$

$$\mu_{011} = \kappa_{011} + \kappa_{010}\kappa_{001}$$

$$\mu_{111} = \mu_{111} + \kappa_{100}\kappa_{011} + \kappa_{010}\kappa_{101} + \kappa_{110}\kappa_{001} + \mu_{100}\mu_{010}\mu_{001}$$

Local moments and cumulants

- 1 Take a box $B(\epsilon) = \prod_{i=1}^P [\xi_i - \frac{\epsilon}{2}]$,
- 2 Multivariate Taylor expansion of the pdf, f , at (ξ_1, ξ_2, \dots)
- 3 Truncate and normalise the expansion over $B(\epsilon)$ (care needed)
- 4 Compute moments and cumulants
- 5 Let $\epsilon \rightarrow 0$
- 6 Investigate the dominant terms
- 7 Result: square free cumulants dominate!

Theorems

Theorem (Local moments)

Let X in \mathbb{R}^p be an absolutely continuous random vector with density f_X which is p times differentiable in ξ in \mathbb{R}^p . Let k in \mathbb{N}^p determine the order of moment. Then, for $|A|$ sufficiently small, X has local moment

$$m_k^A = r(\epsilon, k) \left(\frac{D^\alpha f_X(\xi)}{f_X(\xi)} + O(\epsilon^2) \right), \quad (1)$$

where $r(\epsilon, k) := e^{\|k\|_1^+} \prod_{\substack{i=1, \\ k_i \in 2\mathbb{N}}}^p \frac{1}{k_i+1} \prod_{\substack{i=1, \\ k_i \in 2\mathbb{N}+1}}^p \frac{1}{k_i+2}$ and $\alpha := \sum_{\substack{i=1, \\ k_i \in 2\mathbb{N}+1}}^p e_i$.

Corollary (Local cumulants)

$$\kappa_k^A = \sum_{\pi \in \Pi(k)} \mathbf{c}(\pi) (-1)^{(|\pi|-1)} (|\pi|-1)! \prod_{j=1}^{|\pi|} r(\epsilon, \nu_{M_j}) \left(\frac{D^{\alpha_j} f_X(\xi)}{f_X(\xi)} + O(\epsilon^2) \right),$$

where α_j is a function of the partition π and defined as

$$\alpha_j := \sum_{i=1}^p \mathbf{e}_i \mathbf{1} \left(\nu_{M_j}(i) \in 2\mathbb{N} + 1 \right),$$

that is, α_j is binary and holds ones corresponding to odd elements of ν_{M_j} . Furthermore,

$$r(\epsilon, \nu_{M_j}) := \epsilon^{\|\nu_{M_j}\|_1^+} \prod_{\substack{i=1, \\ \nu_{M_j}(i) \in 2\mathbb{N}}}^p \frac{1}{\nu_{M_j}(i) + 1} \prod_{\substack{i=1, \\ \nu_{M_j}(i) \in 2\mathbb{N}+1}}^p \frac{1}{\nu_{M_j}(i) + 2}.$$

Differential cumulants

Definition (Differential cumulant)

For an index vector k in \mathbb{N}^p , the differential cumulant in a in \mathbb{R}^p is defined as

$$\kappa_k^a := \sum_{\pi \in \Pi(k)} c(\pi) (-1)^{(|\pi|-1)} (|\pi|-1)! \prod_{i=1}^{|\pi|} m_{\nu_{M_i}}^a.$$

Lemma (Differential cumulant)

For a differential cumulant in ξ in \mathbb{R}^p of order k in \mathbb{N}^p it holds that

$$\kappa_k^\xi = D^\alpha \log(f_X(\xi)),$$

where $\alpha := \sum_{\substack{i=1, \\ k_i \in 2\mathbb{N}+1}}^p e_i$ projects odd elements of k onto one even elements of k onto zero.

Independence and conditional independence via differential cumulants

Proposition (Independence in the bivariate case)

Let X in \mathbb{R}^2 . Then $X_1 \perp X_2 \iff \kappa_{11}^X = 0$ for all x in \mathbb{R}^2 .

Proposition (Conditional independence of two random variables)

Let X in \mathbb{R}^p . Then

$$X_i \perp X_j | X_{-ij} \iff \kappa_k^X = 0 \quad \text{for all } x \text{ in } \mathbb{R}^p,$$

where

$$X_{-ij} := (X_1, \dots, X_{i-1}, X_{i+1}, \dots, X_{j-1}, X_{j+1}, \dots, X_p)$$

and $k = e_i + e_j$, $(i, j) \in \{1, \dots, p\}^2$, $i \neq j$.

Multivariate conditional independence

Proposition (Multivariate conditional independence)

Given three index sets I, J, K which partition $\{1, \dots, p\}$, let $S = \{e_i + e_j, i \in I, j \in J\}$. Then

$$X_I \perp\!\!\!\perp X_J \mid X_K \iff \kappa_k^X = 0 \text{ for all } k \in S \text{ and for all } x \text{ in } \mathbb{R}^p.$$

Hierarchical models

Definition

Given a simplicial complex \mathcal{S} over an index set $\mathcal{N} = \{1, \dots, p\}$ and an absolutely continuous random vector X a hierarchical model for the joint distribution function $f_X(\mathbf{x})$ takes the form:

$$f_X(\mathbf{x}) = \exp \left\{ \sum_{J \text{ in } \mathcal{S}} h_J(\mathbf{x}_J) \right\},$$

where $h_J : \mathbb{R}^J \rightarrow \mathbb{R}$ and \mathbf{x}_J in \mathbb{R}^J is the canonical projection of \mathbf{x} in \mathbb{R}^p onto the subspace associated with the index set J .

Exponential family hierarchical models: parameters which appear in a linear way, (log-linear models)

Hierarchical models via differential cumulants

Theorem

Given a simplicial complex \mathcal{S} on an index set \mathcal{N} , a model g is hierarchical, based on \mathcal{S} if and only if all differential cumulants on the complementary complex vanish everywhere, that is

$$\kappa_K^x = 0, \text{ for all } x \text{ in } \mathbb{R}^p \text{ and for all } K \text{ in } \bar{\mathcal{S}}.$$

Monomial ideals

- An ideal: $\langle g(x), \dots, g_m(x) \rangle$ is the set of all polynomials:

$$s_1(x)g(x) + \dots + s_m(x)g_m(x)$$

- A monomial ideal: all the $g_j(x)$ are monomials. Diagram

⋮	⋮	⋮	⋮	⋮	⋮	
x_2^5	$x_1 x_2^5$	$x_1^2 x_2^5$	$x_1^3 x_2^5$	$x_1^4 x_2^5$	$x_1^5 x_2^5$...
x_2^4	$x_1 x_2^4$	$x_1^2 x_2^4$	$x_1^3 x_2^4$	$x_1^4 x_2^4$	$x_1^5 x_2^4$...
x_2^3	$x_1 x_2^3$	$x_1^2 x_2^3$	$x_1^3 x_2^3$	$x_1^4 x_2^3$	$x_1^5 x_2^3$...
x_2^2	$x_1 x_2^2$	$x_1^2 x_2^2$	$x_1^3 x_2^2$	$x_1^4 x_2^2$	$x_1^5 x_2^2$...
x_2	$x_1 x_2$	$x_1^2 x_2$	$x_1^3 x_2$	$x_1^4 x_2$	$x_1^5 x_2$...
1	x_1	x_1^2	x_1^3	x_1^4	x_1^5	...

- $\therefore \langle x_1 x_2^4, x_1^3 x_2^2, x_1^5 \rangle$

Stanley-Reisner ideal and duality

- Let \mathcal{S} be a hierarchical model simplicial complex, eg cliques: $\{13, 23\}$ (conditional independence).
- Take all facets NOT in \mathcal{S} : $\{12, 123\}$
- Construct the corresponding monomial ideal (Stanley-Reisner)

$$I_{\mathcal{S}} = \langle x_1 x_2 \rangle$$

- Duality (Seidenberg nullstellensatz)

$$D^{110} g(x_1, x_2, x_3) \longleftrightarrow \langle x_1 x_2 \rangle$$

- One-one correspondence

hierarchical models \longleftrightarrow square free monomial ideals

Examples

$$\text{CI} \quad \{13,23\} \longleftrightarrow \langle x_1 x_2 \rangle$$

$$\text{3-cycle} \quad \{12,23,13\} \longleftrightarrow \langle x_1 x_2 x_3 \rangle$$

$$\text{4-cycle} \quad \{12,23,34,14\} \longleftrightarrow \langle x_1 x_3, x_2 x_4 \rangle$$

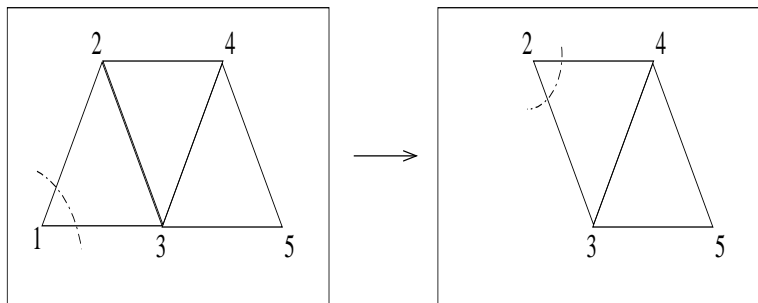
$$\text{decomp} \quad \{123,234,345\} \longleftrightarrow \langle x_1 x_4, x_1 x_5, x_2 x_5 \rangle$$

Decomposability

Definition

Let $\mathcal{N} = \{1, \dots, p\}$ be the vertex set of a graph \mathcal{G} and I, J vertex sets such that $I \cup J = \mathcal{N}$. Then \mathcal{G} is decomposable if and only if $I \cap J$ is complete and I forms a maximal clique or the subgraph based on I is decomposable and similarly for J .

Example of decomposability



Decomposability and marginalisation

Important point: the decomposition of f_{2345} at stage 2 requires a marginalisation.

$$f_V(x_V) = \frac{\prod_{J \in C} f_J(x_J)}{\prod_{K \in S} f_K(x_K)},$$

Two stages

$$f_{12345} = \frac{f_{123} f_{2345}}{f_{23}},$$

$$f_{12345} = \frac{f_{123} f_{234} f_{345}}{f_{23} f_{34}}.$$

Minimal free resolution theorem

- LCM for monomial ideals:

$$x_1 x_2 x_3 \wedge x_4 x_3 x_5 = x_1 x_2 x_3 x_4 x_5$$

- Resolution: monomial maps between successive “levels”, forming an exact sequence (see alg top)
- Minimal free resolution: maps have minimal rank
- Length of MFR: projective dimension
- Linear resolution: all matrices in the resolution have linear terms
- 2-linear: linear and every generator of I_S is of degree 2 (simple interactions)

Alexander duality

- Take the model simplicial complex S .
- Construct the S-R ideal I_S
- The Alexander dual (in monomial form) is all complements of terms in I_S . eg if $n = 5$ the complement of $x_1 x_3 x_5$ is $x_2 x_4$

Example:

$$S = \{123, 234\}$$

Non-faces:

$$\{14, 124, 134, 1234\}$$

Complements:

$$\{\emptyset, 2, 3, 23\}$$

Dirac's Theorem

For a model based on a graph G the following are equivalent

- 1 G is chordal
- 2 I_S has a 2-linear resolution
- 3 The projective dimension of I_{S^*} is 1.

Counter example

Model, S : $\{123, 124, 134, 234, 235, 15\}$

Stanley-Reisner ideal, I_S : $\langle 45.125, 135, 1234 \rangle$

$$0 \longrightarrow S \longrightarrow_C S^4 \longrightarrow_B S^4 \longrightarrow_A S \longrightarrow 0$$

$$A = [45, 125, 135, 1234]$$

$$B = \begin{pmatrix} 0 & -12 & -13 & -123 \\ 3 & 4 & 0 & 0 \\ -2 & 0 & 4 & 0 \\ 0 & 0 & 0 & 5 \end{pmatrix}$$

$$C = -[-4, 4, -2, 0]^T$$

$$AB = 0, \quad BC = 0$$

Not 2-linear \Rightarrow not decomposable

Ferrer ideals

	6	7	8	9
1	$x_1 x_6$	$x_1 x_7$	$x_1 x_8$	
2	$x_2 x_6$	$x_2 x_7$		
3	$x_3 x_6$	$x_3 x_7$		
4	$x_4 x_6$			
5	$x_5 x_6$			

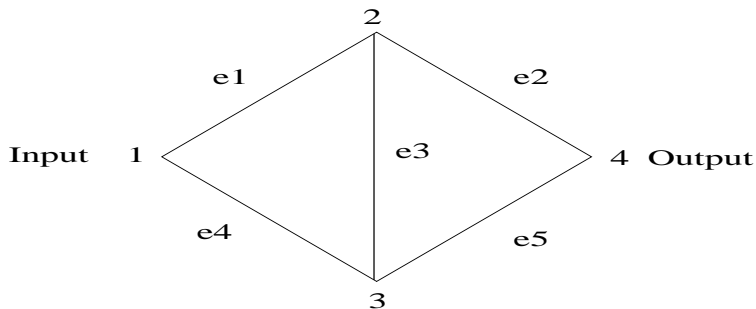
	6	7	8	9
1				$x_1 x_9$
2			$x_2 x_8$	$x_2 x_9$
3			$x_3 x_8$	$x_3 x_9$
4		$x_4 x_7$	$x_4 x_8$	$x_4 x_9$
5		$x_5 x_7$	$x_5 x_8$	$x_5 x_9$

- Ferrer is 2-linear
- Take Ferrer as the Stanley-Reisner ideal $I_{\mathcal{S}}$
- The model \mathcal{S} is decomposable
- Construction: work down the rows of the complementary table.
- Max cliques for the example

$\{123459, 234589, 34589, 45789, 5789\}$

Interpretation in statistics?

Network ideals



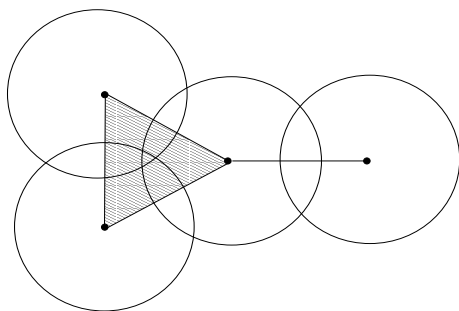
Cut ideal: $I_S = \langle x_1 x_4, x_2 x_5, x_1 x_3 x_5, x_2 x_3 x_4 \rangle$

Path ideal: $I_{S^*} = \langle x_1 x_2, x_4 x_5, x_1 x_3 x_5, x_2, x_3 x_5 \rangle$

Alexander duality: $I_{S^*} = \bar{I}_S$

Extend to “all terminal” reliability: generators (for “paths”) are all minimal spanning trees. Cuts are multipartite graphs.

Persistent homology constructions



- Simplicial complex depends on the centres of the sphere and the radius
- Nerve of the cover has the same topology as cover (Borsuk)
- Delauney complex \cap nerve, has same topology (Naiman and W)
- Different metrics, different types of cover
- Building models using persistent homology ideas

Polynomials and Artinian closure

Lemma

Impose additional differential conditions of the form

$$\frac{\partial^{n_i}}{\partial x_i^{n_i}} g(\mathbf{x}) = 0, \quad \text{for all } 1 \leq i \leq p \text{ and } n \in \mathbb{N}^p \quad (2)$$

the h -functions in the corresponding hierarchical model are polynomials, in which the degree of x_i does not exceed $n_i - 1$, for all $1 \leq i \leq p$.

- Set all $D^\alpha g = 0$ with $|\alpha| = 3$, gives quadratic. Add a NND conditions gives Gaussian
- Set all

$$\frac{\partial^2}{\partial x_i^2} g = 0$$

gives MEC: multivariate exponential class.

More general differential closure?

- How do we deal with other non-polynomials models?
- Are there any interesting continuous exponential family graphical models outside: exponential and Gaussian?
- Yes: eg multivariate von Mises:

$$f = \exp(h\text{-functions})$$

h -functions are terms like

$$\cos(x), \sin(2x), \cos(2x + y), \sin(x + y), \dots$$

But cos and sin satisfy differential equations

Conclusion: new classes of distributions

$$f = \exp(\text{Weyl, D-modules})$$

Parameters appear naturally as constants of integration

- Big problem: closed form for normalising constant (partition/potential function). Takayama et al.

Shellability

Many ideal properties for I_S need to be investigated for their implications for the model S . Shellability is one. Hibbi and Herzog (2011).

- Basic idea: we can build up a complex by adding cliques, of the same max dimension in a special order
- The “join” has one less dimension, but need not be a simplex
- Weaker than decomposability
- But has some decomposability hidden inside
- Can be generalised
- Related to other properties: Cohen-Macaulay, projective dimension, Krull dimension

Further work

- Testing the cumulant condition: DB PhD theseis.
- Connect decomposability to ideal properties
- The key construction is the MFR: make more use of it
- Betti numbers: ordinary, graded, multigraded
- Building models by “growing” simplicial complexes
- Interpretation of “interactions”