

A TERMORDING FREE VARIATION OF MÖLLER ALGORITHM

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DEGRÖBNERIZATION

This conference is part of a series of articles and conferences in the context of Degröbnerization prepared with Michela Ceria, Samuel Lundqvist and Andrea Visconti.

Degröbnerization was introduced for the first time in 2010 into a course at Trento's Cryptolab, implicitly in a commutative setting, but later explicitly in a non-commutative settings at ACA2018 and UMI2019 and was definitely formalized in a conference at ACA2021.

DEGRÖBNERIZATION

Gröbner bases's theory plays an important role in Computer Algebra and many applications have been solved by considering them as a preprocessing, and saying "if we have the Gröbner basis, then the problem is easily solved". This is undoubtedly true, but it does not take into account that *finding a Gröbner basis is not always an easy task*. The computation can become computationally hard and there are cases in which it is even infeasible.

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Degröbnerization consists change perspective in the algebraic representation of our problems, substituting the prior representation, based on *polynomial ideals* to a representation given by *quotient algebras*, expressed via a vector-space basis and multiplication (Auzinger-Stetter) matrices

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requires at most the evaluation of each such functional to each term needed to express the wanted vector-space basis .

SETTING

$$\mathcal{T} := \{x_1^{\gamma_1} \cdots x_n^{\gamma_n} \mid \gamma_1, \dots, \gamma_n \in \mathbb{N}\} \subset \mathcal{S} = \langle x_1, \dots, x_n \rangle$$

ordered by a total ordering (not necessarily a semigroup one)

$\mathcal{P} := \mathbf{k}[x_1, \dots, x_n] = \text{Span}_{\mathbf{k}} \mathcal{T}$ and $\mathcal{Q} := \mathbf{k}\langle x_1, \dots, x_n \rangle = \text{Span}_{\mathbf{k}} \mathcal{S} \supseteq \mathcal{P}$.

An effective ring \mathfrak{A} given as a \mathbf{k} -submodule of either \mathcal{P} or \mathcal{Q} via an ordered subset \mathcal{U} of terms of either \mathcal{T} or \mathcal{S}

A finite (not necessarily linearly independent) set

$\mathbb{L} = \{L_i, 1 \leq i \leq s\} \subset \text{Hom}_{\mathbf{k}}(\mathcal{U}, \mathbf{k})$ of \mathbf{k} -linear functionals $L_i : \mathcal{U} \rightarrow \mathbf{k}$

AIM

For $I := \{f \in \mathcal{U} : L_i(f) = 0\}$ our combinatorial tools return an order ideal \mathbf{N} such that $\mathcal{U}/I \cong \text{Span}_k \mathbf{N}$ where

$$\#(\mathbf{N}(I)) = \deg(I) = \dim_k(\mathbb{L}) =: r \leq s = \#\mathbb{L}.$$

AIM

For $I := \{f \in \mathcal{U} : L_i(f) = 0\}$ produce

an integer $r \in \mathbb{N}$,

an order ideal $\mathbf{N} := \{t_1, \dots, t_r\} \subset \mathcal{T}$,

an ordered subset $\Lambda := \{\lambda_1, \dots, \lambda_r\} \subset \mathbb{L}$

an ordered set $\mathbf{q} := \{q_1, \dots, q_r\} \subset \mathcal{P}$

such that it holds:

$$r = \deg(I) = \dim_k(\mathbb{L}),$$

$$\mathbf{N}(I) = \mathbf{N},$$

$$\text{Span}_k(\Lambda) = \text{Span}_k(\mathbb{L}),$$

$$\text{Span}_k\{t_1, \dots, t_\sigma\} = \text{Span}_k\{q_1, \dots, q_\sigma\}, \forall \sigma \leq r,$$

$$\{q_1, \dots, q_\sigma\}, \{\lambda_1, \dots, \lambda_\sigma\} \text{ are triangular, } \forall \sigma \leq r.$$

TOOL

$\mathbb{L} = \{L_i, 1 \leq i \leq s\} \subset \text{Hom}_{\mathbf{k}}(\mathcal{U}, \mathbf{k}), \mathbb{I} := \{f \in \mathcal{U} : L_i(f) = 0\}$

$\mathbf{N}(\mathbb{I}) = \{t_1, \dots, t_r\}$

Consider the $s \times r$ matrix $\ell_i(t_j)$ whose columns are the vectors $v(t_j, \mathbb{L})$ and are linearly independent, since any relation $\sum_j c_j v(t_j, \mathbb{L}) = 0$ would imply

$$\ell_i\left(\sum_j c_j t_j\right) = \sum_j c_j \ell_i(t_j) = 0 \text{ and } \sum_j c_j t_j \in \{f \in \mathcal{U} : L_i(f) = 0\} = \mathbb{I}$$

contradicting the definition of $\mathbf{N}(\mathbb{I})$.

The matrix $\ell_i(t_j)$ has rank $r \leq s$ and it is possible to extract an ordered subset

$$\Lambda := \{\lambda_1, \dots, \lambda_r\} \subset \mathbb{L}, \quad \text{Span}_{\mathbf{k}}\{\Lambda\} = \text{Span}_{\mathbf{k}}\{\mathbb{L}\}$$

and to re-enumerate the terms in $\mathbf{N}(\mathbb{I})$ in such a way that each principal minor $\lambda_i(t_j)$, $1 \leq i, j \leq \sigma \leq r$ is invertible.

TOOL:BORDER

The *border* of \mathbf{N} is the set of terms $\mathbf{B} := \{x_i t : t \in \mathbf{N}\} \setminus \mathbf{N}$. With this notation, the related border bases are the sets $\mathcal{B} = \{t - \mathbf{Nf}(t) : t \in \mathbf{B}\}$ and $\mathcal{A} := \{t - \mathbf{Nf}(t) : t \in \mathbf{B} \cup \mathbf{N}\}$ where $\mathbf{Nf}(t)$ is the normal form of t , the only polynomail $t - \sum_{\tau \in \mathbf{N}(l)} c_\tau \tau \in I$.

PISTONE G., RICCOMAGNO E., ROGANTIN M.P., **METHODS IN ALGEBRAIC STATISTICS FOR THE DESIGN OF EXPERIMENTS IN OPTIMAL DESIGN AND RELATED AREAS IN OPTIMIZATION AND STATISTICS**, 97–132, 2009, SPRINGER

¹Corner cut: a Hierarchical monomial basis which is the normal set-escalier related to a Gröbner basis

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Another class of statistical models we shall consider are linear models whose vector space basis is formed by **polynomials** v_j which are not **monomials**

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Another class of statistical models we shall consider are linear models whose vector space basis is formed by **polynomials** v_j which are not **monomials**. In Example 7 we show that the model $1, x_1, x_1^2, x_2, x_2^2$ is not a **corner cut**¹ model.

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Another class of statistical models we shall consider are linear models whose vector space basis is formed by **polynomials** v_j which are not **monomials**. In Example 7 we show that the model $1, x_1, x_1^2, x_2, x_2^2$ is not a **corner cut**¹ model. However, it is the most symmetric of the models in the statistical fan. In fact, **to destroy symmetry is a feature of Gröbner basis computation**, as term orderings intrinsically do not preserve symmetries, which are often preferred in statistical models

That's why you should never think of using Gröbner bases in Algebraic Statistics

¹Corner cut: a Hierarchical monomial basis which is the normal set-escalier related to a Gröbner basis

EXAMPLE

$\mathcal{P} := \mathbb{Q}[x, y]$.

The design ideal $\mathbb{I}(\mathcal{F})$ with

$$\mathcal{F} = \{(0, 0), (1, -1), (-1, 1), (0, 1), (1, 0)\}$$

and the Hierarchical monomial basis $\{1, x_1, x_1^2, x_2, x_2^2\}$.

EXAMPLE

$$\mathcal{P} := \mathbb{Q}[x, y].$$

Let us consider the five points

$$P_1 = (0, 0), P_2 = (1, -1), P_3 = (-1, 1), P_4 = (0, 1), P_5 = (1, 0)$$

and the related set of functionals $\mathbb{L} = \{L_1, \dots, L_5\}$ such that L_i is the evaluation at P_i , for $i = 1, \dots, 5$.

$$\mathbb{L} = \{L_i, 1 \leq i \leq s\} \subset \text{Hom}_{\mathbb{Q}}(\mathcal{P}, \mathbb{Q}), \mathbb{I} := \{f \in \mathcal{P} : L_i(f) = 0\}$$

$$\mathbf{N}(\mathbb{I}) = \{t_1, \dots, t_r\}.$$

ALGORITHM:OUTPUT

The algorithm is iterative on each point/functional. At the step i it considers the data for the ideal $I_i := \{f \in \mathcal{P} : f(P_j) = 0, 1 \leq j \leq i\}$, take the point P_{i+1} and return the same data for

$$I_{i+1} := \{f \in \mathcal{P} : f(P_j) = 0, 1 \leq j < i\}$$

the associated escalier $\mathbf{N}(\mathbb{L}_5) = \{t_1, t_2, t_3, t_4, t_5\}$;

$$\mathbf{N}(\mathbb{L}_5) = \{t_1 = 1, t_2 = xt_1 = x, t_3 = xt_2 = x^2, t_4 = yt_1 = y, t_5 = yt_4 = y^2\}$$

$$P_1 = (0, 0), P_2 = (1, -1), P_3 = (-1, 1), P_4 = (0, 1)$$

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The algorithm is iterative on each point/functional. At the step i it considers the data for the ideal $I_i := \{f \in \mathcal{P} : f(P_j) = 0, 1 \leq j \leq i\}$, take the point P_{i+1} and return the same data for

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$$P_1 = (0, 0), P_2 = (1, -1), P_3 = (-1, 1), P_4 = (0, 1)$$

the current border bases \mathcal{A}

$$\mathcal{A} = \{v_1 = 1, v_2 = x, v_3 = x^2 - x = xv_2 - v_2, v_4 = y + x =$$

$$yv_1 + v_2, v_5 = y^2 - y - x - x^2, xv_4 - v_4 - v_5/2 = 1/2x^2 + xy - 1/2x + 1/2y^2 - 1/2y = yv_2 + v_2 + v_3 + v_5/2, xv_3 + v_3 =$$

$$x^3 - x, xv_5 - v_5 = -x^3 + xy^2 - xy + x - y^2 + y, yv_3 - v_3 = x^2y - xy - x^2 + x, yv_4 - v_4 = y^2 + xy - x - y, yv_5 = y^3 - x^2y - xy - y^2\}$$

ALGORITHM:OUTPUT

The algorithm is iterative on each point/functional. At the step i it considers the data for the ideal $I_i := \{f \in \mathcal{P} : f(P_j) = 0, 1 \leq j \leq i\}$, take the point P_{i+1} and return the same data for

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the current border bases \mathcal{A} and \mathcal{B}

$$\mathcal{A} = \{v_1 = 1, v_2 = x, v_3 = x^2 - x = xv_2 - v_2, v_4 = y + x =$$

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$$\mathcal{B} = \{x^3 - x, x^2y - 1/2x^2 + 1/2x + 1/2y^2 - 1/2y, y^3 - y, xy - 1/2x + 1/2y^2 - 1/2y + 1/2x^2, xy^2 - 1/2x - 1/2y^2 + 1/2y + 1/2x^2\}.$$

ALGORITHM:OUTPUT

The algorithm is iterative on each point/functional. At the step i it considers the data for the ideal $I_i := \{f \in \mathcal{P} : f(P_j) = 0, 1 \leq j \leq i\}$, take the point P_{i+1} and return the same data for

$$I_{i+1} := \{f \in \mathcal{P} : f(P_j) = 0, 1 \leq j < i\}$$

$$P_1 = (0, 0), P_2 = (1, -1), P_3 = (-1, 1), P_4 = (0, 1)$$

a basis for the quotient algebra modulo I_5 , i.e.

$$V(\mathbb{L}_5) := \{v_1, v_2, v_3\};$$

$$v_1 = 1, v_2 = x, v_3 = x^2 - x = xv_2 - v_2, v_4 = y + x = yv_1 + v_2, v_5 = y^2 - y - x - x^2$$

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$$P_1 = (0,0), P_2 = (1, -1), P_3 = (-1, 1), P_4 = (0, 1)$$

the Auzinger-Stetter matrices;

$$A_x = \begin{array}{r|ccccc} & v_1 & v_2 & v_3 & v_4 & v_5 \\ \hline v_1 = 1 & 0 & 1 & 0 & 0 & 0 \\ v_2 = x & 0 & 1 & 1 & 0 & 0 \\ v_3 = x^2 - x & 0 & 0 & -1 & 0 & 0 \\ v_4 = y + x & 0 & 0 & 0 & 0 & -1/2 \\ v_5 = y^2 - y - x - x^2 & 0 & 0 & 0 & 0 & 1 \end{array},$$

$$A_y = \begin{array}{r|ccccc} & v_1 & v_2 & v_3 & v_4 & v_5 \\ \hline v_1 = 1 & 0 & -1 & 0 & 1 & 0 \\ v_2 = x & 0 & -1 & -1 & 0 & -1/2 \\ v_3 = x^2 - x & 0 & 0 & 1 & 0 & 0 \\ v_4 = y + x & 0 & 0 & 0 & 1 & 0 \\ v_5 = y^2 - y - x - x^2 & 0 & 0 & 0 & 0 & 0 \end{array}$$

ALGORITHM:OUTPUT

The algorithm is iterative on each point/functional. At the step i it considers the data for the ideal $I_i := \{f \in \mathcal{P} : f(P_j) = 0, 1 \leq j \leq i\}$, take the point P_{i+1} and return the same data for

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$$P_1 = (0, 0), P_2 = (1, -1), P_3 = (-1, 1), P_4 = (0, 1)$$

a triangular set \mathfrak{T} of polynomials;

$$\mathfrak{T} = \{w_1 = v_1 = 1, w_2 = v_2 = x, w_3 = v_3/2 = (x^2 - x)/2, w_4 = v_4 = y + x, w_5 = -v_5/2 = -\frac{y^2 - y - x - x^2}{2}\}$$

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$$P_1 = (0, 0), P_2 = (1, -1), P_3 = (-1, 1), P_4 = (0, 1)$$

$$\begin{aligned} \text{a separator family } \mathfrak{S} = \{ & s_1 = v_1 - v_2 - v_3 - v_4 - v_5/2 = \\ & -1/2 * x^2 - (1/2)x - (1/2)y^2 - (1/2)y + 1, s_2 = \\ & v_2 + v_3/2 + v_5/2 = (1/2)y^2 - (1/2)y, s_3 = v_3/2 = \frac{x^2-x}{2}, s_4 = \\ & v_4 + v_5/2 = -(1/2)x^2 + (1/2)x + (1/2)y^2 + (1/2)y, s_5 = \\ & -v_5/2 = -\frac{y^2-y-x-x^2}{2} \} \end{aligned}$$

$$\Lambda := \{\lambda_1, \dots, \lambda_r\} \subset \mathbb{L} = \{L_i, 1 \leq i \leq s\}$$

Finally **column reduction** allows to linearly express in terms of Λ the elements $L_i \in \mathbb{L} \setminus \Lambda$

	L_1 (0,0)	L_2 (1,0)	L_3 (-1,0)	L_4 (0,1)	L_5 (1,1)	L_6 (-1,1)	L_7 (0,-1)	L_8 (1,-1)	L_9 (-1,-1)
v_1	1	1	1	1	1	1	1	1	1
v_2	0	1	-1	0	1	-1	0	1	-1
v_3	0	0	1	0	0	1	0	0	1
v_4	0	0	0	1	1	1	-1	-1	-1
v_5	0	0	0	0	1	-1	0	-1	1
v_6	0	0	0	0	0	1	0	0	-1

$$0 = L_7 + L_4 - 2L_1 = L_8 + L_5 - 2L_2 = L_9 + L_6 - 2L_3$$

WARNING: CONNECTED-TO-1

Let us consider $\mathbf{k}[x, y]$ and denote $t = xy$. All terms of this ring will be uniquely represented in the form $x^i t^j y^k : ik = 0$.

Thus $x^i t^j y^k \leq x^a t^b y^c$ if

$$i + k < a + c \text{ or}$$

$$i + k = a + c \text{ and } j < b \text{ or}$$

$$i + k = a + c \text{ and } j = b \text{ and } i > c.$$

So that we have $1 < xy < x^2 y^2 < \dots < t^j < t^{j+1} < \dots < x < x^2 y < \dots < x^{j+1} y^j < \dots < y < xy^2 < \dots < x^j y^{j+1} < \dots$

Let us consider the following point set

$$\mathbf{X} = \{(1, 0), (0, 1), (1, 1), (0, 0)\}.$$

WARNING: CONNECTED-TO-1

$$P_1 = (1, 0), P_2 = (0, 1), P_3 = (1, 1), P_4 = (0, 0)$$

	L_1 (1, 0)	L_2 (0, 1)	L_3 (1, 1)	L_4 (0, 0)
$v_1 = x$	1			
$x^2 - x$	0	0	0	0
xy	0	0		
$v_2 = y$	0	1		
$y^2 - y$	0	0	0	0
$v_3 = xy$	0		1	
$x^2y - xy$	0	0	0	0
$xy^2 - xy$	0	0	0	0

CONNECTED TO E

Let $\mathcal{V} \subset \mathcal{P}$ and denote

$$\mathcal{V}^+ := \{\bar{v}_0 + \sum_{i=1}^n x_i \bar{v}_i, \bar{v}_i \in \mathcal{V}, 0 \leq i \leq n\},$$

for each $d \in \mathbb{N} \setminus \{0\}$ set $\mathcal{V}^{[d]} = (\mathcal{V}^{[d-1]})^+$ starting from

$$\mathcal{V}^{[0]} = \text{Span}_{\mathbf{k}}(\mathcal{V}),$$

$$\mathcal{V}^{[*]} := \bigcup_{d \geq 0} \mathcal{V}^{[d]}$$

$\mathcal{V}^{[*]}$ coincides with the ideal generated by \mathcal{V} .

DEFINITION (MOURRAIN)

A vector space $\mathcal{V} \subset \mathcal{P}$ is said to be *connected* to $\bar{e} \in \mathcal{V}$ if, denoting $\mathcal{E} := \text{Span}_{\mathbf{k}}\{\bar{e}\}$, for each $\bar{v} \in \mathcal{V} \setminus \mathcal{E}$, there exists $l > 0$ such that $\bar{v} \in \mathcal{E}^{[l]}$ and $\bar{v} = \bar{v}_0 + \sum_{i=1}^n x_i \bar{v}_i$, with $\bar{v}_i \in \mathcal{E}^{[l-1]} \cap \mathcal{V}$, $0 \leq i \leq n$. \square

CONNECTED TO E

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THEOREM

If \mathcal{V} is connected to \bar{e} , each element of \mathcal{V} satisfies any property if it is satisfied by \bar{e} and

it is satisfied by each linear combination of elements on \mathcal{V} which satisfies it.

THEOREM

If \mathcal{V} is connected to \bar{e} , each element of \mathcal{V} satisfies any property if it is satisfied by \bar{e} and

it is satisfied by each linear combination of elements on \mathcal{V} which satisfies it.

Assume that the property is satisfied by each $\bar{v} \in \mathcal{E}^{[l-1]}$ and let $\bar{w} \in \mathcal{E}^{[l]}$. Since \mathcal{V} is connected to \bar{e} , we have $\bar{w} = \bar{v}_0 + \sum_{i=1}^n x_i \bar{v}_i$, $\bar{v}_i \in \mathcal{E}^{[l-1]}$, $0 \leq i \leq n$. and, by linearity, the property is satisfied also by $\bar{w} \in \mathcal{E}^{[l]}$. The claim then follows by induction.

WARNING: CONNECTED-TO-1

$P_1 = (1, 0), P_2 = (0, 1), P_3 = (1, 1), P_4 = (0, 0)$ $E = \{x, y\}$ $\mathbf{1}$ is not connected to E whose elements vanish in P_4

A finite (not necessarily linearly independent) set

$$\mathbb{L} = \{L_i, 1 \leq i \leq s\} \subset \text{Hom}_{\mathbf{k}}(\mathcal{P}, \mathbf{k}) \dim_{\mathbf{k}}(\mathbb{L}) = r \leq s$$

An ordered subset $\mathcal{U} = \{u_1, \dots, u_r\}$ of terms of either \mathcal{T} or \mathcal{S} ,

$\mathcal{U} = r$ such that $\mathbf{1}$ be connected to \mathcal{U}

$$\mathbf{1} = \sum_{j=1}^r a_j u_j$$

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$$\bar{w} = \bar{v}_0 + \sum_{h=1}^n x_h \bar{v}_h, \bar{v}_h = \sum_{j=1}^r a_{hj} u_j$$

$$\implies \bar{w} = \sum_{j=1}^r \left(a_{0j} + \sum_{h=1}^n a_{hj} \sum_{l=1}^r \alpha_{hlj} \right) v_j$$

Thank you!