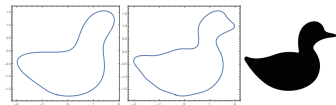


On Moment Problems with Holonomic Functions

F. Bréhard, M. Joldes, J-B. Lasserre



Moment Problems and Applications

Moments of a measure

$$m_\alpha = \int_{\mathbb{R}^n} x^\alpha d\mu \quad \text{for } \alpha \in \mathbb{N}^{n^a}$$

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- n -dim (compact) semi-algebraic set G , with $g \in \mathbb{K}[x]$ vanishing on ∂G
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- **Direct problem**: knowing G and f , find a *complete* system of recurrences for (m_α)
- ↪ **Finite determinacy** of such measures
 - ↪ Solved with **Creative Telescoping**, e.g., [Oaku2013] + Takayama’s algorithm

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- **Inverse problem**: reconstruct G and/or f , given finitely many moments m_α
- Applications in Statistics, Signal-processing, Combinatorics...

Inverse problem: Data recovery from moments

Reconstruction of a shape $G \subset \mathbb{R}^n$ (convex or not)

from the knowledge of finitely many moments

$$m_\alpha = \int_G x^\alpha dx, \quad |\alpha| \leq N,$$

for some given integer N , is a challenging problem.

*Lasserre, J.-B., Putinar M., *Algebraic-exponential Data Recovery from Moments.*, Discrete & Comp Geometry 54.4 (2015): 993-1012.

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A Toulouse Duck?

$m[0,0] = 0.25371$; $m[0,1] = 0.10738$; $m[1,0] = 0.13670$; $m[0,2] = 0.05205$; $m[1,1] = 0.06143$; $m[2,0] = 0.08248$; ...

$$m_{i,j} = \int_{\text{Duck}} x^i y^j dx dy$$



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- Many algorithms developed in optimization, analysis or statistics*
 - Numerical methods, e.g.: convex polytopes [GolubMilanfarVarah1999] [GravinLasserrePasechnikRobins2012]; planar quadrature domains [EbenfeltEtAl2005]; sublevel set of homogeneous polynomials [Lasserre2013]; shape and Gaussian Mixture reconstruction [diDioKummer2019]

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 - Symbolic/algebraic methods:

Multivariate extensions of Prony's method, e.g. [Mourrain2018] (Reconstruction of sparse exponential functions ($\sum_{\alpha \in I} \lambda_\alpha e^{\alpha x}$) from evaluations, moments of Dirac measures); reconstruction of **univariate** piecewise D-finite densities [Batenkov2009]

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Exact Support and/or Density reconstruction

- Lasserre and Putinar's exact reconstruction algorithm (2015)

Theorem 1 (Inverse Problem: Lebesgue measure, Algebraic support)

Let $G \subset \mathbb{R}^n$, bounded open set, whose algebraic boundary is included in the zero set of a polynomial $g \in \mathbb{K}[x]_d$. Given d and a finite number of power moments m_α , up to order $|\alpha| = 3d$, the coefficients of g can be exactly recovered.

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- generalization in the framework of holonomic distributions

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Our contribution: a computer algebra approach

- generalization in the framework of holonomic distributions
- exact recovery of both support and Exp-Polynomial density $f = \exp(p)$, with explicit bound on the required number of moments
- similar algorithm for holonomic density, but no a priori bound on the required number of moments

Table of Contents

Holonomic Distributions and Recurrences on Moments

Inverse Problem: Algorithms and Proofs

- Exponential-Polynomial Densities
- The General Case with D-Finite Densities

Limits and Perspectives

Operator Algebras, Differential Equations and Recurrences

Differential equations/recurrences are translated to skew polynomials:

1. Differential Ore Algebras

- $\mathbb{K}[x]\langle\partial_x\rangle$ **polynomial** Ore algebra

$$\partial_{x_i} x_i = x_i \partial_{x_i} + 1$$

- $\mathbb{K}(x)\langle\partial_x\rangle$ **rational** Ore algebra

$$\partial_{x_i} q(x) = q(x) \partial_{x_i} + \frac{\partial q(x)}{\partial x_i}, \quad q(x) \in \mathbb{K}(x)$$

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$\Rightarrow f$ is **D-finite** iff
 $\mathbb{K}(x)\langle\partial_x\rangle/\mathfrak{Ann}(f)$ has **finite**
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Example:

$$f(x) = c \exp(p(x)) \quad \text{with} \quad p \in \mathbb{K}_s[x]$$

$$f'_{x_i} - p'_{x_i} f = 0$$

$\Rightarrow \mathfrak{Ann}(f)$ generated by the $\partial_{x_i} - p'_{x_i}$

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$$\Rightarrow x \in \mathfrak{Ann}(\delta), \text{ but } 1 \notin \mathfrak{Ann}(\delta)$$

$$\Rightarrow \delta \text{ is } \mathbf{holonomic}$$

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Ore Algebras, Differential Equations and Recurrences

2. Difference Ore Algebras

- Difference operators: **non-commutative**, spanned by $\alpha_1, S_{\alpha_1}, \dots, \alpha_n, S_{\alpha_n}$

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Goals

Recurrences for the moments $m_\alpha = \int_G x^\alpha f(x) dx$:

- **Direct problem:** $\mathfrak{I} \subseteq \mathfrak{Ann}(f) \xrightarrow{?} \mathfrak{J} \subseteq \mathfrak{Ann}(m_\alpha)$
- **Inverse problem:** Reconstruct G and $\mathfrak{I} \subseteq \mathfrak{Ann}(f)$ from sufficiently many m_α

Holonomic Measures

- Measure $\mu = f\mathbf{1}_G$ as a linear functional:

$$\langle f\mathbf{1}_G, \varphi \rangle = \int_{\mathbb{R}^n} \varphi(x)f(x)\mathbf{1}_G(x)\mathrm{d}x = \int_G \varphi(x)f(x)\mathrm{d}x$$

- Action of Ore polynomials: $L\mu = ?$

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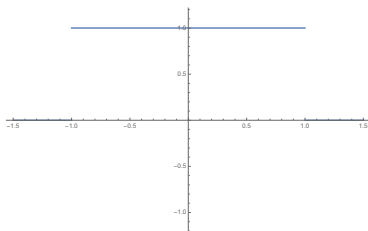
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Example: Lebesgue measure over a segment

Let $G = [-1, 1]$, $f = 1$, and $\mu = \mathbf{1}_G$

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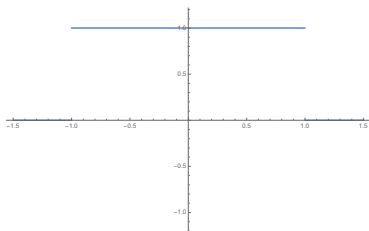
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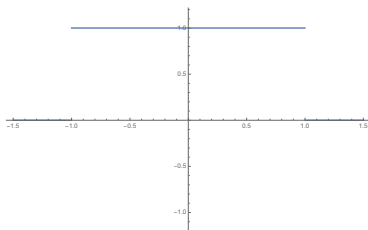
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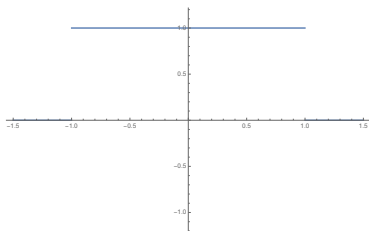
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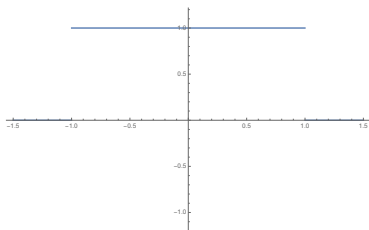
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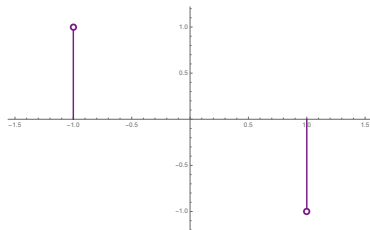
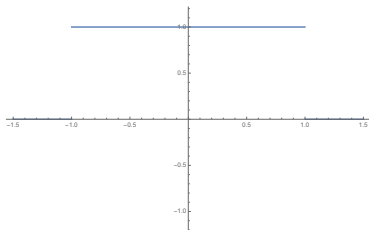
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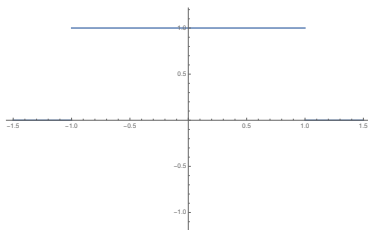
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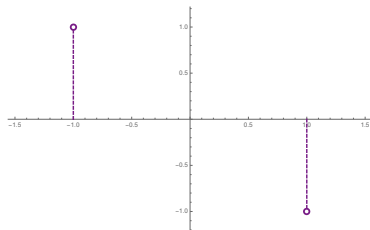
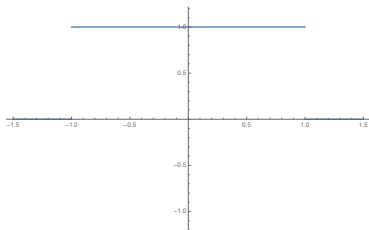
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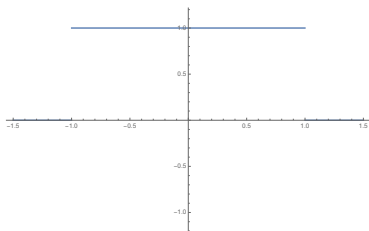
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Holonomic Measures

- Measure $\mu = f\mathbf{1}_G$ as a linear functional:

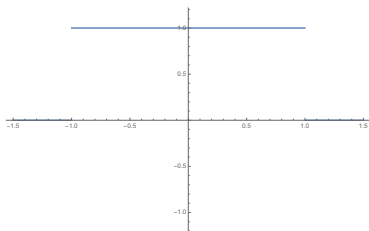
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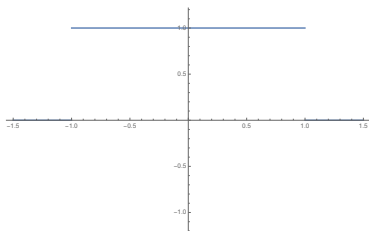
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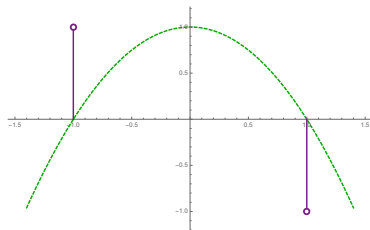
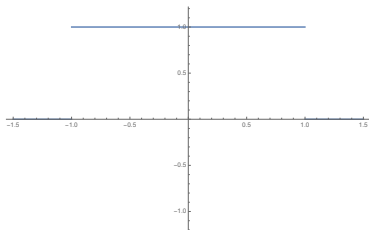
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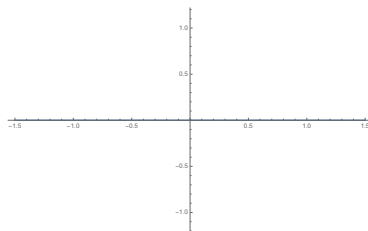
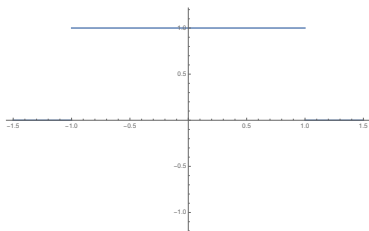
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- $\mathfrak{Ann}(T)$ in $\mathbb{K}[x] \langle \partial_x \rangle \Rightarrow$ **holonomic** instead of D-finite

From Holonomic Measures to Recurrences on Moments

Example: Lebesgue measure over a segment (continued)

Let $G = [-1, 1]$, $f = 1$, $\mu = \mathbf{1}_G$ and $\varphi = x^k$:

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\Rightarrow Recurrence satisfied by the moments (m_k) :

$$(k+2) m_{k+1} - k m_{k-1} = 0$$

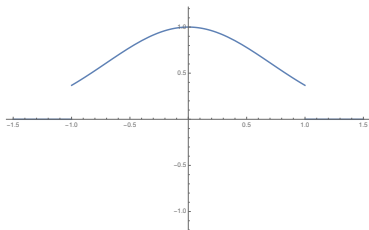
This is indeed true...

$$m_k = \int_{-1}^1 x^k dx = \begin{cases} \frac{2}{k+1} & \text{if } k \text{ even} \\ 0 & \text{if } k \text{ odd} \end{cases}$$

Using Integration by Parts

Example: Exp-Poly density over a segment

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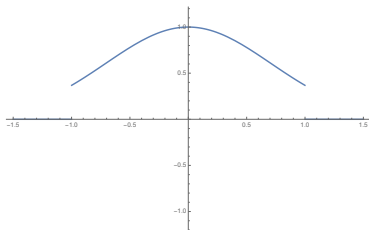


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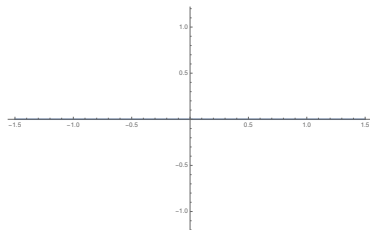
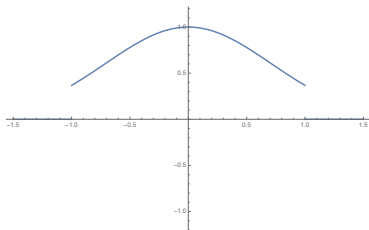


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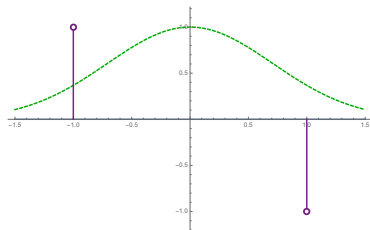
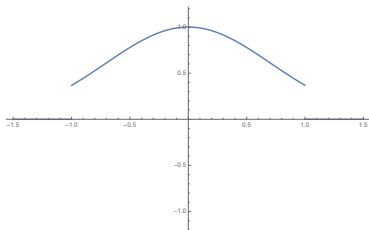


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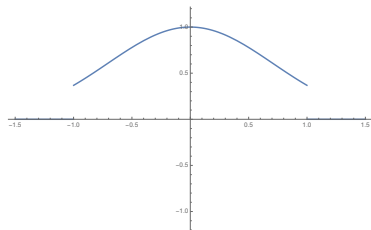


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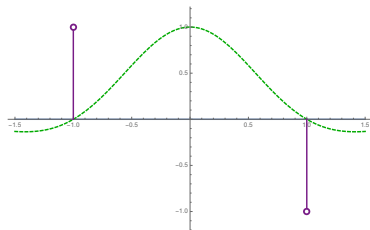
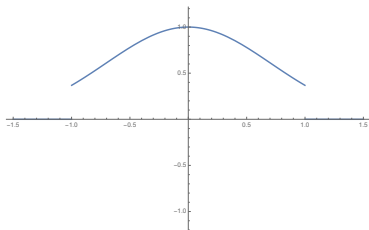


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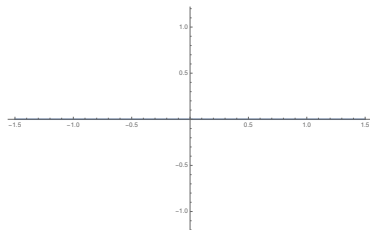
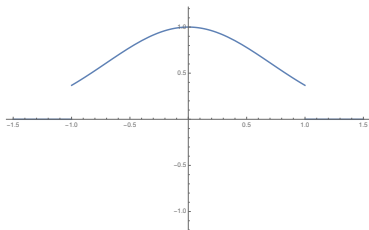
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The General Case

$$\mu = f \mathbf{1}_G, \quad L \in \mathbb{K}[x] \langle \partial_x \rangle \text{ of order } r,$$

- Use **Lagrange identity**:

$$\varphi(Lf) - (L^* \varphi)f = \partial_x \mathcal{L}_L(f, \varphi)$$

→ \mathcal{L}_L bilinear concomitant in f, φ with derivatives of order $\leq r - 1$

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$$\Rightarrow \bar{L} = g^r L \in \mathfrak{Ann}(\mu)$$

From Differential Equations to Recurrences

- Translate $\bar{L} = g^r L \in \mathfrak{Ann}(\mu)$ into a recurrence on (m_α) :

$$x_i \rightarrow S_{\alpha_i} \qquad \partial_{x_i} \rightarrow -\alpha_i S_{\alpha_i}^{-1}$$

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Direct Problem

1. $\{L_1, \dots, L_k\} \subseteq \mathfrak{Ann}(f)$ D-finite
2. $\{\bar{L}_1, \dots, \bar{L}_k\} \subseteq \mathfrak{Ann}(\mu)$
3. Translate into $\{R_1, \dots, R_k\} \subseteq \mathfrak{Ann}(m_\alpha)$
4. Gröbner basis algo on $\{R_1, \dots, R_k\}$

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 $\{x \in \mathbb{C}^n \mid g(x) = 0 \text{ and } \nabla g(x) = 0\} = \emptyset$,
 then the recurrences system is holonomic.

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◦ Reconstruct \bar{L}_i , then g and L_i from the given moments m_α

\Rightarrow Translation $\bar{L}_i \leftrightarrow R_i$ is linear

Note: Actual proof of holonomicity of $\{R_1, \dots, R_k\}$ **not** needed

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- The General Case with D-Finite Densities

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Inverse Problem — Roadmap and Issues

— To reconstruct g vanishing on ∂G and $L \in \mathfrak{Ann}(f)$ of order r :

1. Make an **ansatz** \tilde{L} for $\bar{L} = g^r L \in \mathfrak{Ann}(\mu)$
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$$\langle \tilde{L} \mu, x^\alpha \rangle = \langle \mu, \tilde{L}^* x^\alpha \rangle = \int_G (\tilde{L}^* x^\alpha) f(x) dx = 0, \quad |\alpha| \leq N \quad (LS_N)$$

requiring moments m_α for $|\alpha| \leq N + \dots$

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— Issues to be handled:

- **False** solutions in (LS_N) : $\tilde{L} \notin \mathfrak{Ann}(\mu)$?
- How many moments m_α : **a priori bounds** on N ?
- Can g and L be always extracted from $\tilde{L} \in \mathfrak{Ann}(\mu)$?

Reconstruction of Exp-Poly Densities

— $\mu = f\mathbf{1}_G$ with $f(x) = \exp(p(x))$ for $p \in \mathbb{K}[x]_s$ and $g \in \mathbb{K}[x]_d$ vanishing on ∂G

$$\overline{L}_i = g(\partial_{x_i} - p'_{x_i}) \in \mathfrak{Ann}(\mu)$$

Reconstruction of Exp-Poly Densities

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Input: Moments m_α of μ for $|\alpha| \leq N + d + s - 1$

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Theorem — Correctness of RECONSTRUCTEXPOLY

If $N \geq 3d + s - 1$, then RECONSTRUCTEXPOLY computes:

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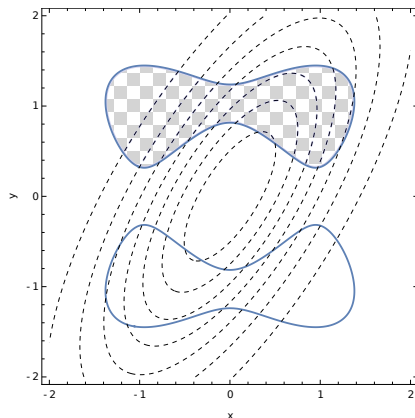
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Example — Algebraic Support, Gaussian Measure

→ Reconstruction of:

$$f(x, y) = \exp(-x^2 + xy - y^2/2) \quad \text{and} \quad g(x, y) = (x^2 - 9/10)^2 + (y^2 - 11/10)^2 - 1$$

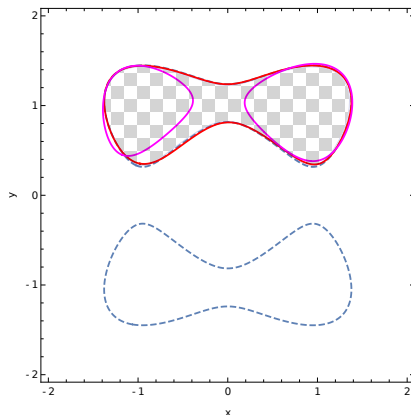


Moments $(m_{ij})_{i+j \leq 18}$ with **10** digits of accuracy

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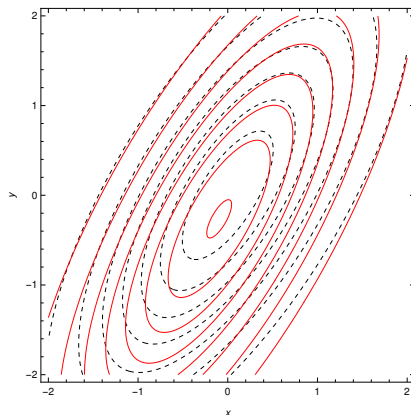


Moments $(m_{ij})_{i+j \leq 18}$ with **4, 6, 8** digits of accuracy

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Moments $(m_{ij})_{i+j \leq 18}$ with 8 digits of accuracy

Density and Support Reconstruction in the General Case

— $\mu = f\mathbf{1}_G$ with $g \in \mathbb{K}[x]_d$ vanishing on ∂G , and $\{L_1, \dots, L_n\}$ **rectangular** system for f :

$$L_i = q_{ir_i} \partial_{x_i}^{r_i} + \dots + q_{i1} \partial_{x_i} + q_{i0} \in \mathfrak{Ann}(f) \cap \mathbb{K}[x] \langle \partial_{x_i} \rangle$$

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1. Compute coefficients of ansatz $\tilde{h} \in \mathbb{K}[x]_{dr}$ with nontrivial solution of

$$\langle \mu, (\tilde{h} L_i)^* x^\alpha \rangle = 0, \quad 1 \leq i \leq n, \quad |\alpha| \leq N$$

2. $\tilde{g}^r \leftarrow \tilde{h}$ $(\tilde{g} \leftarrow \tilde{h} / \text{GCD}(\tilde{h}, \tilde{h}'_{x_1}, \dots, \tilde{h}'_{x_n}))$

Density and Support Reconstruction in the General Case

Theorem — Correctness of RECONSTRUCTDENSITY

For N large enough, the rectangular system $\{\tilde{L}_1, \dots, \tilde{L}_n\}$ computed by RECONSTRUCTDENSITY is in $\mathfrak{Ann}(f)$.

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Theorem — Correctness of RECONSTRUCTSUPPORT

RECONSTRUCTSUPPORT computes $\tilde{g} = \lambda g$ with $\lambda \neq 0$ whenever $q_{ir} \neq 0$ on ∂G and $N \geq (2r - 1)d + (d - 1)b + s$ where:

- $r = \max_{1 \leq i \leq n} r_i$, orders of the L_i
- $b = r \bmod 2$
- $s = \max_{1 \leq i \leq n} \{\deg(q_{ir})\}$ maximal degree of the head coefficients

Support Reconstruction — Proof

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- Suppose that $\tilde{h} = g^k h_0$ with $g \nmid h_0$ and $k < r$

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$$\begin{aligned} \mathcal{L}_{L_i}(f, \tilde{h} \varphi) &= f \left[q_{i1} \tilde{h} \varphi - \partial_{x_i}(q_{i2} \tilde{h} \varphi) + \cdots + (-1)^{r-1} \partial_{x_i}^{r-1}(q_{ir} \tilde{h} \varphi) \right] \\ &\quad + \partial_{x_i}(f) \left[q_{i2} \tilde{h} \varphi - \partial_{x_i}(q_{i3} \tilde{h} \varphi) + \cdots + (-1)^{r-2} \partial_{x_i}^{r-2}(q_{ir} \tilde{h} \varphi) \right] \\ &\quad + \cdots \\ &\quad + \partial_{x_i}^{r-1}(f) q_{ir} \tilde{h} \varphi. \end{aligned}$$

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⇒ **Contradiction:** $h_0 = 0$ on ∂G , hence $g \mid h_0$

The Singular Case — Example in Combinatorics

→ Express **Catalan numbers** as moments of a measure μ :

$$C_n = \frac{1}{n+1} \binom{2n}{n} \stackrel{?}{=} \int_I x^n f(x) dx$$

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$$C_n = \lambda \int_{-\infty}^{+\infty} x^n \sqrt{\frac{4-x}{x}} dx \quad ?$$

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$$C_n = \frac{1}{2\pi} \int_0^4 x^n \sqrt{\frac{4-x}{x}} dx \quad ?$$

Table of Contents

Holonomic Distributions and Recurrences on Moments

Inverse Problem: Algorithms and Proofs

- Exponential-Polynomial Densities
- The General Case with D-Finite Densities

Limits and Perspectives

Some Limits and Perspectives

- A priori bounds for N in the general case with unknown D-finite density?
- Full determination of the density, including initial conditions
- Extracting the component of $V(g)$ corresponding to ∂G

Bounds for the Number of Moments?

- Is there an explicit bound N_0 on N s.t. for ansatz \tilde{L} of $\bar{L} = g^r L$:

$$\langle \tilde{L} \mu, \varphi \rangle = 0 \quad \text{for all } \varphi \in \mathbb{K}[x]_N \quad \Rightarrow \quad \tilde{L} \mu = 0 \quad \text{when } N \geq N_0 \quad ?$$

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- Such a bound N_0 depending only on the structure of \tilde{L} cannot exist:

Example [Batenkov2009] — Legendre Polynomials P_n over $[-1, 1]$

$P_n(x)$ annihilated by $L_n = (1 - x^2)\partial_x^2 - 2x\partial_x + n(n+1) \Rightarrow$ common ansatz \tilde{L}

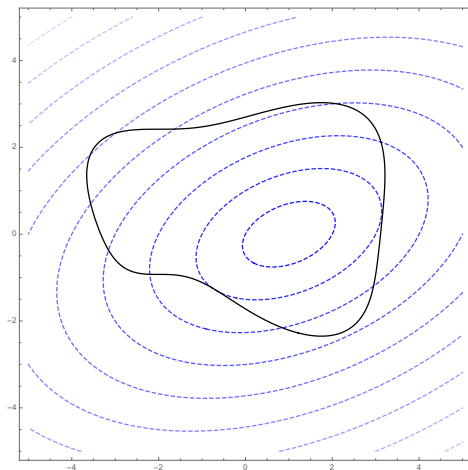
but $m_k^{(n)} = \int_{-1}^1 x^k P_n(x) dx = 0 \quad \text{for } k < n \quad \text{and} \quad m_n^{(n)} > 0$

- Explicit bounds depending on upper bounds on the coefficients of \tilde{L} ?

Reconstructing Initial Conditions of the Density

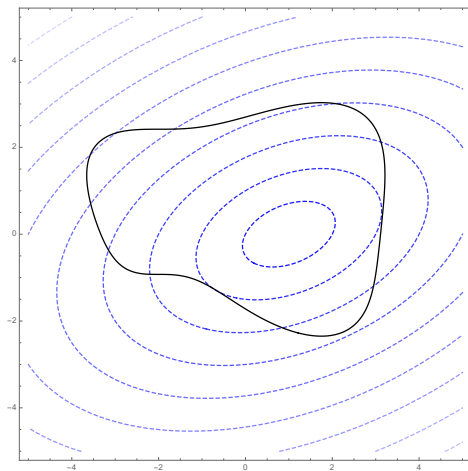
- Algorithm RECONSTRUCTDENSITY only computes a system $\tilde{\mathcal{I}} = \{\tilde{L}_1, \dots, \tilde{L}_n\}$ but not the initial conditions that fully characterize f

Reconstructing Initial Conditions of the Density



$$f(x, y) = \lambda_1 e^{p_1(x, y)}$$
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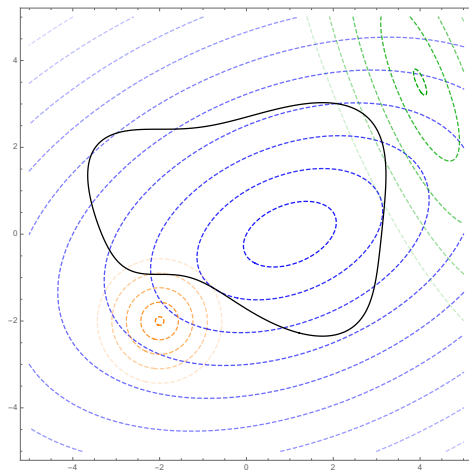
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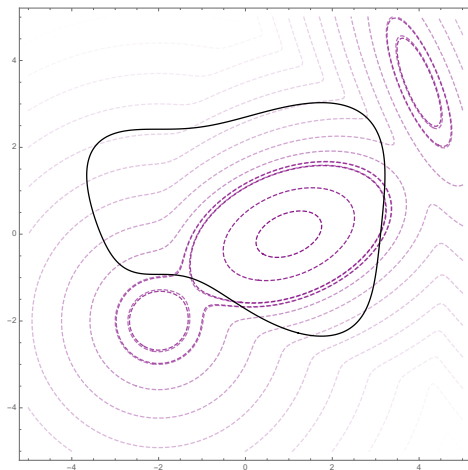
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Reconstructing Initial Conditions of the Density



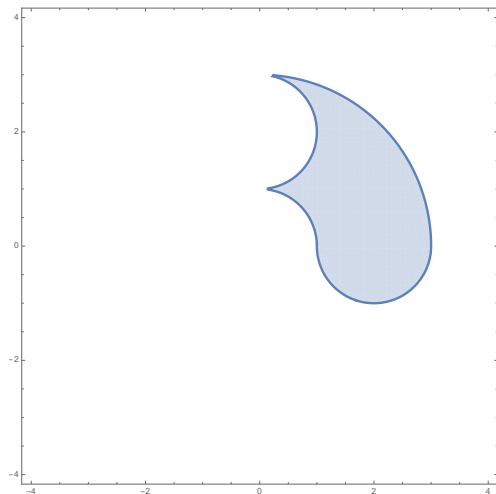
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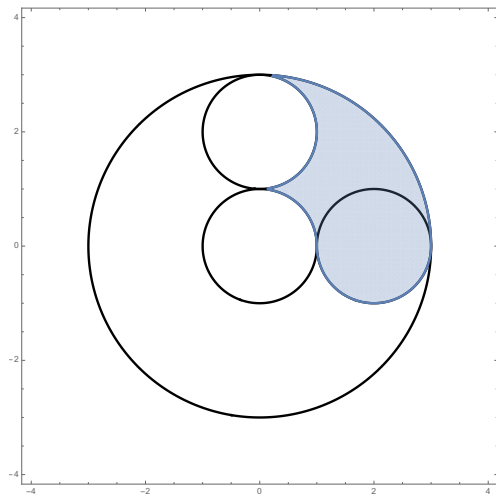
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- Algorithm RECONSTRUCTDENSITY only computes a system $\tilde{\mathcal{I}} = \{\tilde{L}_1, \dots, \tilde{L}_n\}$ but not the initial conditions that fully characterize f
- Solution: compute initial moments for a basis of solution densities of $\tilde{\mathcal{I}}$
 - Optimization techniques, e.g., [HenrionLasserreSavorgnan2009]
 - Computer algebra, e.g., [LairezMezzarobbaElDin2019]

Isolation of the Topological Boundary

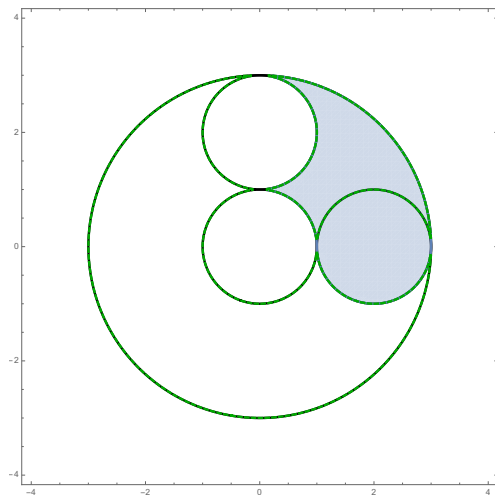


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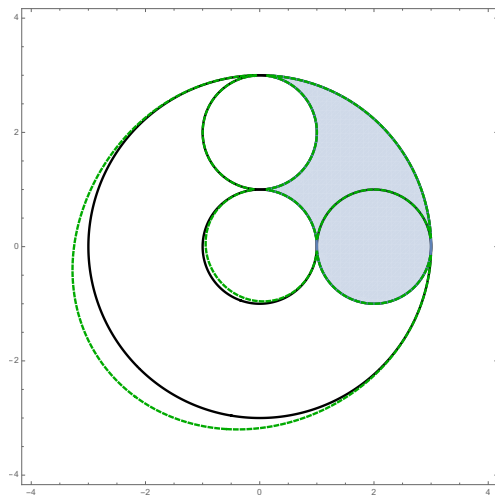
$$I(\partial G) = (g) \quad \text{with} \quad g(x, y) = (x^2 + y^2 - 9)(x^2 + y^2 - 1)((x - 2)^2 + y^2 - 1)(x^2 + (y - 2)^2 - 1)$$

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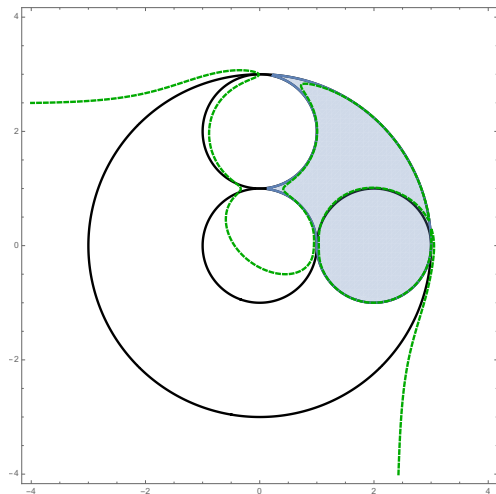
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 $\textcolor{green}{\tilde{g}}$ reconstructed using $\textcolor{brown}{6}$ digits accuracy for the moments (m_α)

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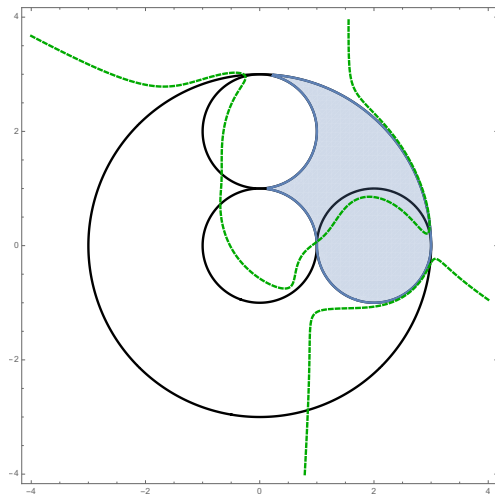
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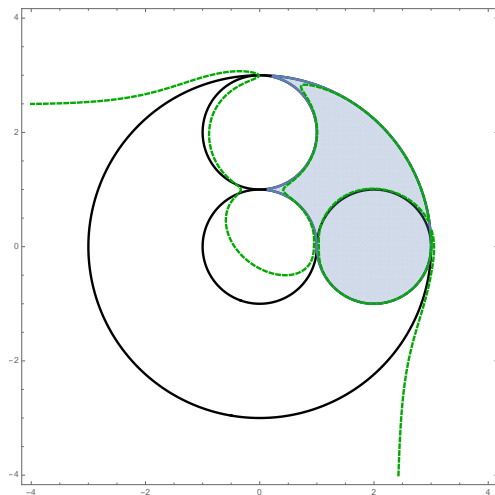
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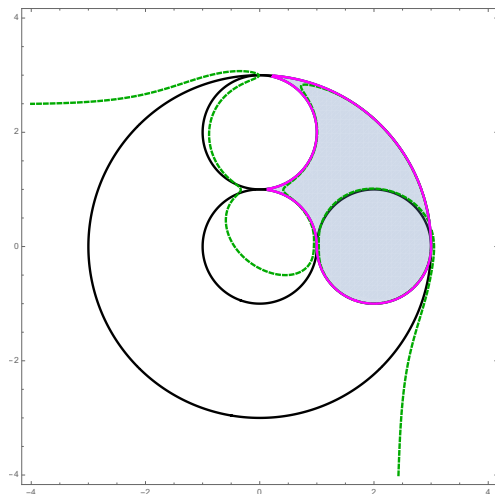
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$$\partial \mathbf{G} \approx \{(x, y) \mid g(x, y) = 0 \text{ and } \mathbb{E}[\tilde{g}(x, y)^2] \leq \epsilon\}, \quad \tilde{g} \leftarrow \text{randomly perturbed } (m_\alpha)$$

Conclusion and Perspectives

Contributions:

- Extension of [LasserrePutinar2015] to reconstruction of unknown **Exp-Poly** density and unknown semi-algebraic support
- Explicit bound for the number N of required moments
- Reconstruction algorithm for unknown **holonomic** density and unknown semi-algebraic support
- Numerical experiments using **least-squares** approximation when approximate moments are known

Future work:

- Generic bounds for N depending on the magnitude of the coefficients
- **Numerical** aspects: robustness w.r.t. approximate moments, or nonpolynomial boundary
- Isolation of the **topological** boundary via perturbation techniques
- Application to problems involving **combinatorial sequences**

Thank you for your attention!