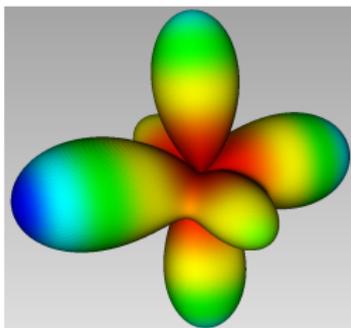


From moments to sparse representations, a geometric, algebraic and algorithmic viewpoint

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Part II



- 1 **Decomposition algorithms**
- 2 Low rank classification
- 3 The variety of missing moments

Univariate series:



Kronecker (1881)

The Hankel operator

$$\begin{aligned}
 H_\sigma : \mathbb{C}^{\mathbb{N}, \text{finite}} &\rightarrow \mathbb{C}^{\mathbb{N}} \\
 (p_m) &\mapsto \left(\sum_m \sigma_{m+n} p_m \right)_{n \in \mathbb{N}}
 \end{aligned}$$

is of **finite rank** r iff $\exists \omega_1, \dots, \omega_{r'} \in \mathbb{C}[y]$ and $\xi_1, \dots, \xi_{r'} \in \mathbb{C}$ distincts s.t.

$$\sigma(y) = \sum_{n \in \mathbb{N}} \sigma_n \frac{y^n}{n!} = \sum_{i=1}^{r'} \omega_i(y) \mathbf{e}_{\xi_i}(y)$$

with $\sum_{i=1}^{r'} (\deg(\omega_i) + 1) = r$.

Multivariate series:

Theorem (Generalized Kronecker Theorem)

For $\sigma = (\sigma_1, \dots, \sigma_m) \in (R^*)^m$, the Hankel operator

$$\begin{aligned} H_\sigma : R &\rightarrow (R^*)^m \\ p &\mapsto (p \star \sigma_1, \dots, p \star \sigma_m) \end{aligned}$$

is of rank r iff

$$\sigma_j = \sum_{i=1}^{r'} \omega_{j,i}(\mathbf{y}) \mathbf{e}_{\xi_i}(\mathbf{y}) \in \mathcal{P}olExp, \quad j = 1, \dots, m$$

with $r = \sum_{i=1}^{r'} \mu(\omega_{1,i}, \dots, \omega_{m,i})$. In this case, we have

- ▶ $\mathcal{V}_{\mathbb{C}}(I_\sigma) = \{\xi_1, \dots, \xi_{r'}\}$.
- ▶ $I_\sigma = Q_1 \cap \dots \cap Q_{r'}$ with $Q_i^\perp = \langle\langle \omega_{1,i}, \dots, \omega_{m,i} \rangle\rangle \mathbf{e}_{\xi_i}(\mathbf{y})$.

If $m = 1$, \mathcal{A}_σ is **Gorenstein** ($\mathcal{A}_\sigma^* = \mathcal{A}_\sigma \star \sigma$ is a free \mathcal{A}_σ -module of rank 1) and $(a, b) \mapsto \langle \sigma | ab \rangle$ is non-degenerate in \mathcal{A}_σ .

The decomposition from the algebraic structure

Decomposition problem

Given a (truncated) sequence of moments σ_α , $\alpha \in A$, find $\xi_j = (\xi_{j,1}, \xi_{j,1}, \dots, \xi_{j,n}) \in \overline{\mathbb{K}}^n$ distinct, $\omega_j \in \overline{\mathbb{K}}$. s.t. $\sigma = \sum_j \omega_j e_{\xi_j}$

Hankel operator: For $\sigma \in R^*$,

$$\begin{aligned} H_\sigma : R &\rightarrow R^* \\ p &\mapsto p \star \sigma \end{aligned}$$

Quotient algebra: $\mathcal{A}_\sigma = R/I_\sigma$ where $I_\sigma = \ker H_\sigma$.

$$\begin{aligned} 0 \rightarrow I_\sigma \rightarrow \mathbb{K}[\mathbf{x}] &\xrightarrow{H_\sigma} \mathcal{A}_\sigma^* \rightarrow 0 \\ p &\mapsto p \star \sigma \end{aligned}$$

Isomorphism between \mathcal{A}_σ and $\mathcal{A}_\sigma^* = I_\sigma^\perp$.

(\mathcal{A}_σ **Gorenstein**, i.e. $\exists \tau = \sigma \in \mathcal{A}_\sigma^*$ s.t. $\mathcal{A}_\sigma^* = \mathcal{A}_\sigma \star \tau$ is a free \mathcal{A}_σ -module).

👉 Find the points ξ_j as the roots of I_σ and the weights ω_j from the idempotents of \mathcal{A}_σ .

The structure of \mathcal{A}_σ

For $\sigma = \sum_{i=1}^r \omega_i \mathbf{e}_{\xi_i}$, with $\omega_i \in \mathbb{C} \setminus \{0\}$ and $\xi_i \in \mathbb{C}^n$ distinct.

- ▶ rank $H_\sigma = r$ and the multiplicity of the points ξ_1, \dots, ξ_r in $\mathcal{V}(I_\sigma)$ is 1.
- ▶ For B, B' be of size r , $H_\sigma^{B', B}$ invertible iff B and B' are bases of $\mathcal{A}_\sigma = \mathbb{K}[\mathbf{x}]/I_\sigma$.
- ▶ The matrix M_i of multiplication by x_i in the basis B of \mathcal{A}_σ is such that

$$\mathbf{H}_\sigma^{B', x_i B} = \mathbf{H}_{x_i * \sigma}^{B', B} = \mathbf{H}_\sigma^{B', B} \mathbf{M}_i$$

- ▶ The common **eigenvectors** of M_i are (up to a scalar) the Lagrange **interpolation polynomials** \mathbf{u}_{ξ_i} at the points ξ_i , $i = 1, \dots, r$.

$$\mathbf{u}_{\xi_i}(\xi_j) = \begin{cases} 1 & \text{if } i = j, \\ 0 & \text{otherwise} \end{cases} \quad \mathbf{u}_{\xi_i}^2 \equiv \mathbf{u}_{\xi_i}, \quad \sum_{i=1}^r \mathbf{u}_{\xi_i} \equiv 1.$$

- ▶ The common **eigenvectors** of M_i^t are (up to a scalar) the vectors $[B(\xi_i)]$, $i = 1, \dots, r$.

Decomposition algorithm

Input: The first coefficients $(\sigma_\alpha)_{\alpha \in A}$ of the series

$$\sigma = \sum_{\alpha \in \mathbb{N}^n} \sigma_\alpha \frac{\mathbf{y}^\alpha}{\alpha!} = \sum_{i=1}^r \omega_i \mathbf{e}_{\xi_i}(\mathbf{y})$$

- ① Compute bases $B, B' \subset \langle \mathbf{x}^A \rangle$ s.t. that $H^{B',B}$ invertible and $|B| = |B'| = r = \dim \mathcal{A}_\sigma$;
- ② Deduce the tables of multiplications $M_i := (H_\sigma^{B',B})^{-1} H_\sigma^{B',x_i B}$
- ③ Compute the eigenvectors $\mathbf{v}_1, \dots, \mathbf{v}_r$ of $\sum_i l_i M_i$ for a generic $\mathbf{l} = l_1 x_1 + \dots + l_n x_n$;
- ④ Deduce the points $\xi_i = (\xi_{i,1}, \dots, \xi_{i,n})$ s.t. $M_j \mathbf{v}_i - \xi_{i,j} \mathbf{v}_i = 0$ and the weights $\omega_i = \frac{1}{\mathbf{v}_i(\xi_i)} \langle \sigma | \mathbf{v}_i \rangle$.

Output: The decomposition $\sigma = \sum_{i=1}^r \frac{1}{\mathbf{v}_i(\xi_i)} \langle \sigma | \mathbf{v}_i \rangle \mathbf{e}_{\xi_i}(\mathbf{y})$.

Demo

Multivariate Prony method (1)

Let $h(t_1, t_2) = 2 + 3 \mathbf{2}^{t_1} \mathbf{2}^{t_2} - \mathbf{3}^{t_1}$, $\sigma = \sum_{\alpha \in \mathbb{N}^2} h(\alpha) \frac{y^\alpha}{\alpha!} = 2\epsilon_{(1,1)}(y) + 3\epsilon_{(2,2)}(y) - \epsilon_{(3,1)}(y)$.

- Take $B = \{1, x_1, x_2\}$ and compute

$$H_0 := H_\sigma^{B,B} = \begin{bmatrix} h(0,0) & h(1,0) & h(0,1) \\ h(1,0) & h(2,0) & h(1,1) \\ h(0,1) & h(1,1) & h(0,2) \end{bmatrix} = \begin{bmatrix} 4 & 5 & 7 \\ 5 & 5 & 11 \\ 7 & 11 & 13 \end{bmatrix},$$

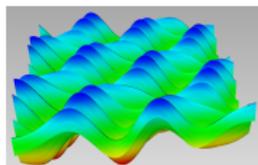
$$H_1 := H_\sigma^{B, x_1 B} = \begin{bmatrix} 5 & 5 & 7 \\ 5 & -1 & 17 \\ 811 & 178 & 23 \end{bmatrix}, \quad H_2 := H_\sigma^{B, x_2 B} = \begin{bmatrix} 7 & 11 & 13 \\ 11 & 17 & 23 \\ 13 & 23 & 25 \end{bmatrix}.$$

- Compute the generalized eigenvectors of $(aH_1 + bH_2, H_0)$:

$$U = \begin{bmatrix} 2 & -1 & 0 \\ -1/2 & 0 & 1/2 \\ -1/2 & 1 & -1/2 \end{bmatrix} \text{ and } H_0 U = \begin{bmatrix} \mathbf{2} & \mathbf{3} & \mathbf{-1} \\ \mathbf{2} \times \mathbf{1} & \mathbf{3} \times \mathbf{2} & \mathbf{-1} \times \mathbf{3} \\ \mathbf{2} \times \mathbf{1} & \mathbf{3} \times \mathbf{2} & \mathbf{-1} \times \mathbf{1} \end{bmatrix}.$$

- This yields the weights $\mathbf{2}, \mathbf{3}, \mathbf{-1}$ and the roots $(\mathbf{1}, \mathbf{1}), (\mathbf{2}, \mathbf{2}), (\mathbf{3}, \mathbf{1})$.

Multivariate Prony method (2)



$$h(t_1, t_2) := \sum_{i=1}^r \omega_i e^{f_1 t_1 + f_2 t_2} \text{ with}$$

$$\mathbf{f} := \begin{bmatrix} -0.5 & 1.0 + 3.141592654 i \\ 0.1 + 21.36283005 i & 1.5 + 32.67256360 i \\ 0.1 + 21.36283005 i & -0.5 + 79.16813488 i \\ -2.5 + 145.7698991 i & -10.0 + 517.1061508 i \end{bmatrix} \quad \omega := \begin{bmatrix} 1.375328890 + 0.9992349291 i \\ 1.046162168 + 0.3399186938 i \\ 0.9 \\ -9.2 \end{bmatrix}$$

For the sampling $[\frac{1}{50}, \frac{1}{170}]$, $B = \{1, x_1, x_2, x_1 x_2\}$, the SVD of $H_\sigma^{B,B}$ is

$$[33.1196344300301391, 14.3767453860219057, 0.244096952193142480, 0.0230734326225932214]$$

and the computed decomposition is

$$\mathbf{f}^* = \begin{bmatrix} -2.49999999703636711 + 145.769899153890435 i & -9.9999999913514852 + 517.106150711515852 i \\ 0.0999940670173818935 + 21.3628392917863437 i & -0.500045063743692286 + 79.1681566575291527 i \\ 0.100028305341504586 + 21.3628527756206275 i & 1.50002358381760881 + 32.6725933609709571 i \\ -0.499926454593063452 + 0.0000142466247443506387 i & 1.00008814016387371 + 3.14161379568963772 i \end{bmatrix}$$

$$\omega^* = \begin{bmatrix} -9.19999999613861696 - 0.00000000772422142913953280 i \\ 0.899999936743709261 - 0.00000156202814849404348 i \\ 1.04615643213670850 + 0.339923495269889020 i \\ 1.37533468654902213 + 0.999231697828891208 i \end{bmatrix}$$

Sparse interpolation

$$f(\mathbf{x}) = \sum_{i=1}^r \omega_i \mathbf{x}^{\alpha_i} \quad \Rightarrow \quad \sigma = \sum_{\gamma} f(\varphi^{\gamma}) \frac{\mathbf{y}^{\gamma}}{\gamma!} = \sum_{i=1}^r \omega_i \mathbf{e}_{\varphi^{\alpha_i}}(\mathbf{y})$$

Example: $f(x_1, x_2) = x_1^{33} x_2^{12} - 5 x_1 x_2^{45} + 101$.

- ▶ Compute $\sigma_{\alpha} = f(\varphi_1^{\alpha_1}, \varphi_2^{\alpha_2})$ for $\alpha_1 + \alpha_2 \leq 3$ and $\varphi_1 = \varphi_2 = e^{\frac{2i\pi}{50}}$.
- ▶ Compute the Hankel matrix $H_{\sigma}^{1,2}$:

$$\begin{bmatrix} 97.00000 & 97.01771 + 3.93695i & 95.50360 - 1.47099i & 98.46280 + 4.88062i & 97.42748 + 1.82098i & 99.50853 + 5.29465i \\ 97.01771 + 3.93695i & 98.46280 + 4.88062i & 97.42748 + 1.82098i & 102.35770 + 3.77300i & 99.50853 + 5.29465i & 95.42134 + 1.47250i \\ 95.50360 - 1.47099i & 97.42748 + 1.82098i & 95.73130 - .33862i & 99.50853 + 5.29465i & 95.42134 + 1.47250i & 99.50853 + 5.29465i \end{bmatrix}$$

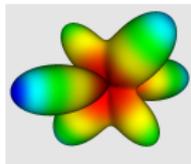
- ▶ Deduce the decomposition of $\sigma = \sum_{i=1}^3 \omega_i \mathbf{e}_{\xi_i}$:

$$\Xi = \begin{bmatrix} 0.99211 + 0.12533i & 0.80902 - 0.58779i \\ 1.00000 + 4.86234e^{-11}i & 1.00000 - 6.91726e^{-10}i \\ -0.53583 - 0.84433i & 0.06279 + 0.99803i \end{bmatrix} \omega = \begin{bmatrix} -5.00000 - 4.43772e^{-7}i \\ 101.00000 + 4.65640e^{-7}i \\ 1.00000 - 1.92279e^{-8}i \end{bmatrix}$$

- ▶ and the exponents $\frac{50\Xi}{2\pi i} \bmod 50$ of the terms of f :

$$\begin{bmatrix} 1.00000 - 0.414119e^{-7}i & -5.00000 + 0.270858e^{-6}i \\ 0.386933e^{-9} + 0.137963e^{-8}i & -0.550458e^{-8} - 0.38761e^{-8}i \\ -17.00000 - 0.100085e^{-6}i & 12.00000 + 0.700984e^{-6}i \end{bmatrix}$$

Symmetric tensor decomposition



$$\begin{aligned}
 \tau &= (\mathbf{x}_0 - \mathbf{x}_1 + 3\mathbf{x}_2)^4 + (\mathbf{x}_0 + \mathbf{x}_1 + \mathbf{x}_2)^4 - 3(\mathbf{x}_0 + 2\mathbf{x}_1 + 2\mathbf{x}_2)^4 \\
 &= -x_0^4 - 24x_0^3x_1 - 8x_0^3x_2 - 60x_0^2x_1^2 - 168x_0^2x_1x_2 - 12x_0^2x_2^2 \\
 &\quad - 96x_0x_1^3 - 240x_0x_1^2x_2 - 384x_0x_1x_2^2 + 16x_0x_2^3 - 46x_1^4 - 200x_1^3x_2 \\
 &\quad - 228x_1^2x_2^2 - 296x_1x_2^3 + 34x_2^4
 \end{aligned}$$

$$\tau^* = e_{(-1,3)}(\mathbf{y}) + e_{(1,1)}(\mathbf{y}) - 3e_{(2,2)}(\mathbf{y}) \quad (\text{by apolarity})$$

$$H_{\tau^*}^{2,2} := \begin{bmatrix} -1 & -2 & -6 & -2 & -14 & -10 \\ -2 & -2 & -14 & 4 & -32 & -20 \\ -6 & -14 & -10 & -32 & -20 & -24 \\ -2 & 4 & -32 & 34 & -74 & -38 \\ -14 & -32 & -20 & -74 & -38 & -50 \\ -10 & -20 & -24 & -38 & -50 & -46 \end{bmatrix}$$

For $B = \{1, x_2, x_1\}$,

$$H_{\tau^*}^{B,B} = \begin{bmatrix} -1 & -2 & -6 \\ -2 & -2 & -14 \\ -6 & -14 & -10 \end{bmatrix}, H_{\tau^*}^{B,x_1B} = \begin{bmatrix} -6 & -14 & -10 \\ -14 & -32 & -20 \\ -10 & -20 & -24 \end{bmatrix}, H_{\tau^*}^{B,x_2B} = \begin{bmatrix} -2 & -2 & -14 \\ -2 & 4 & -32 \\ -14 & -32 & -20 \end{bmatrix}$$

- ▶ The matrix of multiplication by x_1 in $B = \{1, x_2, x_1\}$ is

$$M_1 = (H_{\tau^*}^{B,B})^{-1} H_{\tau^*}^{B,x_1 B} = \begin{bmatrix} 0 & -2 & -2 \\ 0 & \frac{1}{2} & \frac{3}{2} \\ 1 & \frac{5}{2} & \frac{3}{2} \end{bmatrix}.$$

- ▶ Its eigenvalues are $[-1, 1, 2]$ and the eigenvectors:

$$U := \begin{bmatrix} 0 & -2 & -1 \\ \frac{1}{2} & \frac{3}{4} & \frac{1}{2} \\ -\frac{1}{2} & \frac{1}{4} & \frac{1}{2} \end{bmatrix}.$$

that is the polynomials

$$U(x) = \left[\frac{1}{2} x_2 - \frac{1}{2} x_1 \quad -2 + \frac{3}{4} x_2 + \frac{1}{4} x_1 \quad -1 + \frac{1}{2} x_2 + \frac{1}{2} x_1 \right].$$

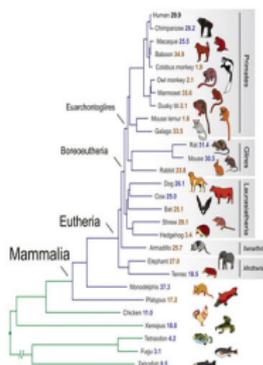
- ▶ We deduce the weights and the frequencies:

$$H_{\tau^*}^{[1, x_1, x_2], U} = \begin{bmatrix} 1 & 1 & -3 \\ 1 \times -1 & 1 \times 1 & -3 \times 2 \\ 1 \times 3 & 1 \times 1 & -3 \times 2 \end{bmatrix}.$$

Weights: $1, 1, -3$; frequencies: $(-1, 3), (1, 1), (2, 2)$.

Decomposition: $\tau^*(\mathbf{y}) = \epsilon_{(-1,3)}(\mathbf{y}) + \epsilon_{(1,1)}(\mathbf{y}) - 3\epsilon_{(2,2)}(\mathbf{y}) + (\mathbf{y})^4$

Phylogenetic trees



Problem: study probability vectors for genes $[A, C, G, T]$ and the transitions described by Markov matrices M^i .

Example:

$$\begin{aligned} \text{Ancestor} &: && \mathcal{A} \\ \text{Transitions} &: && M^1 \quad M^2 \quad M^3 \\ \text{Species} &: && S_1 \quad S_2 \quad S_3 \end{aligned}$$

For $i_1, i_2, i_3 \in \{A, C, G, T\}$, the probability to observe i_1, i_2, i_3 is

$$p_{i_1, i_2, i_3} = \sum_{k=1}^4 \pi_k M_{k, i_1}^1 M_{k, i_2}^2 M_{k, i_3}^3 \Leftrightarrow \mathbf{p} = \sum_{k=1}^4 \pi_k \mathbf{u}_k \otimes \mathbf{v}_k \otimes \mathbf{w}_k$$

where $\mathbf{u}_k = (M_{k,1}^1, \dots, M_{k,4}^1)$, $\mathbf{v}_k = (M_{k,1}^2, \dots, M_{k,4}^2)$, $\mathbf{w}_k = (M_{k,1}^3, \dots, M_{k,4}^3)$.

👉 p is a tensor $\in \mathbb{K}^4 \otimes \mathbb{K}^4 \otimes \mathbb{K}^4$ of rank ≤ 4 .

👉 Its decomposition yields the M^i and the ancestor probability (π_j) .

A general framework

- ▶ \mathfrak{F} the functional space, in which the “signal” lives.
- ▶ $S_1, \dots, S_n : \mathfrak{F} \rightarrow \mathfrak{F}$ commuting linear operators: $S_i \circ S_j = S_j \circ S_i$.
- ▶ $\Delta : h \in \mathfrak{F} \mapsto \Delta[h] \in \mathbb{C}$ a linear functional on \mathfrak{F} .

Generating series associated to $h \in \mathfrak{F}$:

$$\sigma_h(\mathbf{y}) = \sum_{\alpha \in \mathbb{N}^n} \Delta[S^\alpha(h)] \frac{\mathbf{y}^\alpha}{\alpha!} = \sum_{\alpha \in \mathbb{N}^n} \sigma_\alpha \frac{\mathbf{y}^\alpha}{\alpha!}.$$

- ▶ Eigenfunctions:

$$S_j(E) = \xi_j E, j = 1, \dots, n \Rightarrow \sigma_E = \omega \mathbf{e}_\xi(\mathbf{y}).$$

- ▶ Generalized eigenfunctions:

$$S_j(E_k) = \xi_j E_k + \sum_{k' < k} m_{j,k'} E_{k'} \Rightarrow \sigma_{E_k} = \omega_i(\mathbf{y}) \mathbf{e}_\xi(\mathbf{y}).$$

☞ If $h \mapsto \sigma_h$ is injective \Rightarrow unique decomposition of f as a linear combination of generalized eigenfunctions.

Sum of polynomial-exponential functions

- ▶ $\mathcal{F} = \mathcal{PolExp}(\mathbf{x})$,
- ▶ $S_j : h(\mathbf{x}) \mapsto h(x_1, \dots, x_{j-1}, x_j + \delta_j, x_{j+1}, \dots, x_n)$ shift of x_j by δ_j ,
- ▶ $\Delta : h(\mathbf{x}) \mapsto \Delta[h] = h(0)$ the evaluation at 0.

Generating series of h : $\sigma_h(\mathbf{y}) = \sum_{\alpha \in \mathbb{N}^n} h(\alpha_1 \delta_1, \dots, \alpha_n \delta_n) \frac{\mathbf{y}^\alpha}{\alpha!}$.

Eigenfunctions: $e^{\mathbf{f} \cdot \mathbf{x}}$; generalized eigenfunctions: $\omega(\mathbf{x})e^{\mathbf{f} \cdot \mathbf{x}}$;

$h(\mathbf{x}) = \sum_{i=1}^{r'} g_i(\mathbf{x})e^{\mathbf{f}_i \cdot \mathbf{x}} + r(\mathbf{x})$ with $g_i(\mathbf{x}) \in \mathbb{C}[\mathbf{x}]$, $\mathbf{f}_i \in \mathbb{C}^n$ and $r(\delta \odot \alpha) = 0$, $\forall \alpha \in \mathbb{N}^n$, iff

$$\sigma_h(\mathbf{y}) = \sum_{i=1}^{r'} \omega_i(\mathbf{y})e_{\xi_i}(\mathbf{y})$$

with $\xi_i = e^{\mathbf{f}_i} \in \mathcal{V}(\ker H_{\sigma_h}) \subset \mathbb{C}^n$, $\omega_i(\mathbf{x}) = \sum_{\alpha} g_{i,\alpha} \omega_{\alpha}$ for $g_i = \sum_{\alpha} g_{i,\alpha} \mathbf{x}^{\alpha}$.

👉 Decomposition from the moments $\sigma_{\alpha} = h(\alpha_1 \delta_1, \dots, \alpha_n \delta_n)$.

Sparse interpolation of PolyLog functions

- ▶ $\mathcal{F} = \text{PolyLog}(\mathbf{x}) = \{ \sum_{(\beta, \gamma) \in A} h_{\beta, \gamma} \log^\beta(\mathbf{x}) \mathbf{x}^\gamma, A \text{ finite} \},$
- ▶ $S_j : h(x_1, \dots, x_n) \mapsto h(\dots, x_{j-1}, \lambda_j x_j, x_{j+1}, \dots)$ for $\lambda_j \in \mathbb{C} - \{1\},$
- ▶ $\Delta : h(x_1, \dots, x_n) \mapsto \Delta[h] = h(1, \dots, 1).$

Generating series of h : $\sigma_h(\mathbf{y}) = \sum_{\alpha \in \mathbb{N}^n} h(\lambda_1^{\alpha_1}, \dots, \lambda_n^{\alpha_n}) \frac{\mathbf{y}^\alpha}{\alpha!}.$

Eigenfunctions: \mathbf{x}^γ ; generalized eigenfunctions: $\log^\beta(\mathbf{x}) \mathbf{x}^\gamma.$

$h = \sum_{i=1}^{r'} \sum_{\beta \in B_i} \omega_{i, \beta} \log^\beta(\mathbf{x}) \mathbf{x}^{\gamma_i}$ iff the generating series σ_h is of the form

$$\sigma_h(\mathbf{y}) = \sum_{i=1}^{r'} \omega_i(\mathbf{y}) \mathbf{e}_{\xi_i}(\mathbf{y})$$

with $\xi_i = (\lambda_1^{\gamma_{i,1}}, \dots, \lambda_n^{\gamma_{i,n}}) \in \mathbb{C}^n$ and $\omega_i(\mathbf{y}) = \sum_{\beta \in B_i} \omega_{i, \beta} \mathbf{y}^\beta \in \mathbb{C}[\mathbf{y}].$

👉 Decomposition from the moments $\sigma_\alpha = h(\lambda_1^{\alpha_1}, \dots, \lambda_n^{\alpha_n}).$

Sparse reconstruction from Fourier coefficients

- ▶ $\mathcal{F} = L^2(\Omega)$;
- ▶ $S_i : h(x) \in L^2(\Omega) \mapsto e^{2\pi \frac{x_i}{T_i}} h(x) \in L^2(\Omega)$ is the multiplication by $e^{2\pi \frac{x_i}{T_i}}$;
- ▶ $\Delta : h(x) \in \mathcal{O}'_C \mapsto \int h(x) dx \in \mathbb{C}$.

The moments of f are

$$\sigma_\gamma = \frac{1}{\prod_{j=1}^n T_j} \int f(x) e^{-2\pi i \sum_{j=1}^n \frac{\gamma_j x_j}{T_j}} dx$$

Eigenfunctions: δ_ξ ; generalized eigenfunctions: $\delta_\xi^{(\alpha)}$.

For $f \in L^2(\Omega)$ and $\sigma = (\sigma_\gamma)_{\gamma \in \mathbb{Z}^n}$ its Fourier coefficients,

$$\Gamma_\sigma : (\rho_\beta)_{\beta \in \mathbb{Z}^n} \in L^2(\mathbb{Z}^n) \mapsto \left(\sum_{\beta} \sigma_{\alpha+\beta} \rho_\beta \right)_{\alpha \in \mathbb{Z}^n} \in L^2(\mathbb{Z}^n).$$

Γ_σ is of finite rank r if and only if $f = \sum_{i=1}^{r'} \sum_{\alpha \in A_i \subset \mathbb{N}^n} \omega_{i,\alpha} \delta_{\xi_i}^{(\alpha)}$ with $\xi_i = (\xi_{i,1}, \dots, \xi_{i,n}) \in \Omega$, $\omega_{i,\alpha} \in \mathbb{C}$ and $r = \sum_{i=1}^{r'} \mu(\sum_{\alpha \in A_i} \omega_{i,\alpha} \mathbf{y}^\alpha)$

Other applications

- ▶ Decomposition of measures as sums of spikes from moments (images, spectroscopy, radar, astronomy, . . .)
- ▶ Decomposition of convolution operators of finite rank
- ▶ Vanishing ideal of points: $\sigma = \sum_{i=1}^r \epsilon_{\xi_i}(\mathbf{y})$
- ▶ Change of ordering for Grobner bases or change of bases for zero-dimensional ideals: $\sigma_\alpha = \langle u, N(\mathbf{x}^\alpha) \rangle$,
- ▶ . . .

- 1 Decomposition algorithms
- 2 Low rank classification**
- 3 The variety of missing moments

Low rank decomposition of Hankel matrices

Rank 1 Hankel matrices: $H_\xi = [\xi^{\alpha+\beta}]_{\alpha \in A, \beta \in B}$ for some $\xi \in \mathbb{K}^n$ or \mathbb{P}^n .

Rank r Hankel matrices are not necessarily the sum of r rank one Hankel matrices.

$$\begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \neq \lambda_1 \begin{bmatrix} 1 & \xi_1 & \xi_1^2 \\ \xi_1 & \xi_1^2 & \xi_1^3 \\ \xi_1^2 & \xi_1^3 & \xi_1^4 \end{bmatrix} + \lambda_2 \begin{bmatrix} 1 & \xi_2 & \xi_2^2 \\ \xi_2 & \xi_2^2 & \xi_2^3 \\ \xi_2^2 & \xi_2^3 & \xi_2^4 \end{bmatrix}$$

but

$$\begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} = \lim_{\epsilon \rightarrow 0} \frac{1}{2\epsilon} \begin{bmatrix} 1 & \epsilon & \epsilon^2 \\ \epsilon & \epsilon^2 & \epsilon^3 \\ \epsilon^2 & \epsilon^3 & \epsilon^4 \end{bmatrix} - \frac{1}{2\epsilon} \begin{bmatrix} 1 & -\epsilon & \epsilon^2 \\ -\epsilon & \epsilon^2 & -\epsilon^3 \\ \epsilon^2 & -\epsilon^3 & \epsilon^4 \end{bmatrix}$$

Symbol: $y = \lim_{\epsilon \rightarrow 0} \frac{1}{2\epsilon} (e^{\epsilon y} - e^{-\epsilon y})$.

Structured Low rank Decomposition

Decomposition in sum of Hankel operators associated to symbols $\omega(y)\mathbf{e}_\xi(y)$ with $\omega(y) \in \mathbb{K}[\mathbf{y}]$, $\xi \in \mathbb{C}^n$.

$$\Rightarrow \sigma = \sum_{i=1}^r \omega_i \mathbf{e}_{\xi_i}(\mathbf{y}) \Rightarrow$$

$$H_\sigma^{A,B} = V_A(\xi_1, \dots, \xi_r) \Delta(\omega_1, \dots, \omega_r) V_B(\xi_1, \dots, \xi_r)^t$$

$$H_{g^*\sigma}^{A,B} = V_A(\xi_1, \dots, \xi_r) \Delta(\omega_1 g(\xi_1), \dots, \omega_r g(\xi_r)) V_B(\xi_1, \dots, \xi_r)^t$$

where $V_A(\xi_1, \dots, \xi_r) = [\xi_j^{\alpha_i}]_{1 \leq i \leq |A|, 1 \leq j \leq r}$, $\Delta(\dots)$ diagonal matrix.

$$\Rightarrow \sigma = \sum_{i=1}^r \omega_i(\mathbf{y}) \mathbf{e}_{\xi_i}(\mathbf{y}) \Rightarrow$$

$$H_\sigma^{A,B} = W_{A;\Gamma}(\xi) \Delta_\omega^\Gamma W_{B;\Gamma}(\xi)^t$$

$$H_{g^*\sigma}^{A,B} = W_{A;\Gamma}(\xi) \Delta_{g^*\omega}^\Gamma W_{B;\Gamma}(\xi)^t$$

where $W_{A;\Gamma}(\xi)$ Wronskian, $\Delta_{g^*\omega}^\Gamma$ block diagonal.

Symmetric tensors of low rank (joint work with A. Oneto)

For $\psi \in S_d$ of degree d , with a decomposition $\psi = \sum_{i=1}^r \langle \xi_i, \bar{\mathbf{x}} \rangle^d$ and for $0 \leq k \leq d - k$,

$$H_{\psi^*}^{d-k,k} = V_{d-k}(\Xi) V_k^t(\Xi)$$

where $\Xi = (\xi_1, \dots, \xi_r) \in (\mathbb{K}^{n+1})^r$, $V_k(\Xi)$ is the Vandermonde matrix of Ξ at the monomials of deg. k .

Notation

- ▶ $\psi_k^\perp = \ker H_{\psi^*}^{d-k,k}$
- ▶ $h(k) = \dim S_k / \psi_k^\perp = \text{rank} H_{\psi^*}^{d-k,k}$
- ▶ $I(\Xi)$ defining ideal of the points Ξ

Apolarity lemma

Ξ is **apolar** to ψ (i.e. appears in a decomposition of ψ) iff $I(\Xi)_k \subset \psi_k^\perp$ for any $k \in \mathbb{N}$.

Proof. $\psi^* \in \langle \mathbf{e}_{\xi_1}, \dots, \mathbf{e}_{\xi_r} \rangle \subset S_d^*$.

The **regularity** of Ξ is $\rho(\Xi) = \min\{k \in \mathbb{N} \mid \exists u_1, \dots, u_r \in S_k \text{ s.t. } u_i(\xi_j) = \delta_{i,j}\}$.

Regularity lemma

Let $\psi \in S_d$ and let Ξ be a minimal set of points apolar to ψ . Then,
 $I(\Xi)_k = \psi_k^\perp$ for $0 \leq k \leq d - \rho(\Xi)$.

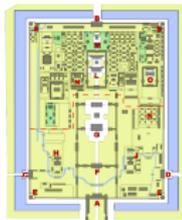
Proof. $\psi_k^\perp = \ker H_{\psi^*}^{d-k,k} = V_{d-k}(\Xi) V_k^t(\Xi)$, $I(\Xi)_k = \ker V_k^t(\Xi)$ and $V_{d-k}(\Xi)$ injective for $d - k \geq \rho(\Xi)$.

Theorem

Let $\psi \in S_d$ and let Ξ be a minimal set of points apolar to ψ . If $d \geq 2\rho(\Xi) + 1$, then

$$I(\Xi) = (\psi_{\leq \rho(\Xi)+1}^\perp).$$

Moreover, Ξ is the unique minimal set of points apolar to ψ .



A set of **essential variables** of ψ is a minimal set of linear forms $l_1, \dots, l_N \in S$, such that $\psi \in \mathbb{C}[l_1, \dots, l_N]$.

Proposition

- ▶ [Car06] the number of essential variables is $h_\psi(1)$;
- ▶ [CC017] any minimal decomposition of ψ involves only linear forms in the essential variables.

The **Waring locus** of ψ is the locus of linear forms that can appear in a minimal decomposition of ψ , i.e.,

$$\mathcal{W}_\psi := \left\{ [l] \in \mathbb{P}(S_1) \mid \exists l_2, \dots, l_r, r = \text{rank}(\psi), \text{ s.t. } \psi \in \langle l^d, l_2^d, \dots, l_r^d \rangle \right\};$$

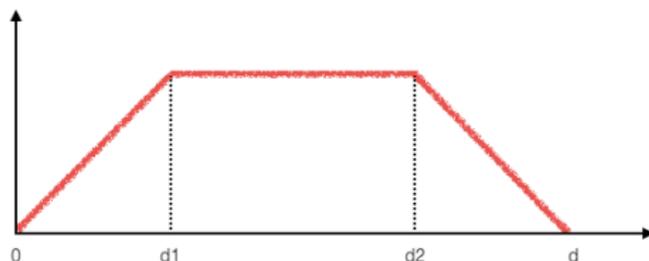
The complement is **forbidden locus** denoted $\mathcal{F}_\psi := \mathbb{P}^n \setminus \mathcal{W}_\psi$.

Tensor with 2 essential variables

(Sylvester method)

Let $\psi(x_0, x_1) \in \mathcal{S}_d = \mathbb{K}[x_0, x_1]_d$.

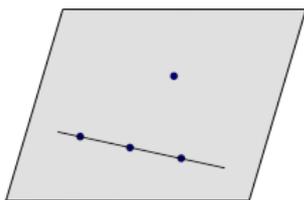
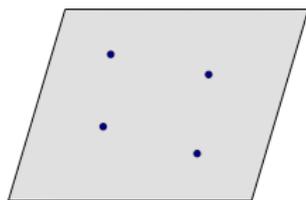
The Hilbert function of \mathcal{A}_{ψ^*} is of the form:



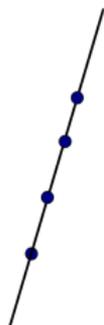
with $(\psi^\perp) = (G_1, G_2)$ of degree $0 \leq d_1 \leq d_2 \leq d$ with $d_1 + d_2 = d + 2$.

- ▶ If G_1 has simple roots, then ψ is of rank $d_1 = \deg(G_1)$ and the roots of G_1 are the unique minimal set apolar to ψ .
- ▶ Otherwise, ψ is of rank $d_2 = \deg(G_2)$ and \mathcal{W}_ψ is dense in \mathbb{P}^1 . For a generic choice of $A \in \mathcal{S}_{d_2-d_1}$, the roots of $AG_1 + G_2$ are a minimal set apolar to ψ .

Cases of rank 4

(a)
Collinear(b) Coplanar,
with 3 collinear.(c) General
coplanar.(d) General
points.

For $\psi \in S_d$ of rank 4.



ψ has two essential variables ($h_\psi(1) = 2$):
 $\psi^\perp = (L_1, \dots, L_{n-1}, G_1, G_2)$, where $\deg(G_i) = d_i$ and $d_1 \leq d_2$. In particular, it has to be $d \geq 4$ and:

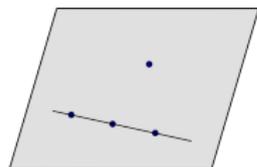
- (i) if $d = 4, 5, 6$, then $d_2 = 4$, and minimal apolar sets of points are defined by ideals $I(\Xi) = (L_1, \dots, L_{n-1}, HG_1 + \alpha G_2)$, for a general choice of $H \in T_{6-d}$ and $\alpha \in \mathbb{C}$;
- (ii) if $d \geq 7$, then $d_1 = 4$ and the unique minimal apolar set of points is given by $I(\Xi) = (L_1, \dots, L_{n-1}, G_1)$.

ψ has three essential variables ($h_\psi(1) = 3$) and a minimal apolar set Ξ of type (b):

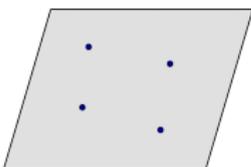
(i) if $d = 3$, then $\mathcal{V}(\psi_2^\perp) = P + D$, where P is a reduced point and D is connected scheme of length 2 whose linear span is a line L_D ;
Any minimal apolar set is of the type $P \cup \Xi'$, with $\Xi' \subset L_D$;

(ii) if $d = 4$, then $h_\psi(2) = 4$, $\mathcal{V}(\psi_2^\perp) = P \cup L$, where P is a reduced point and L is a line not passing through P ;
Any minimal apolar set is of the type $P \cup \Xi'$, where $\Xi' \subset L$.

(iii) if $d \geq 5$, then $h_\psi(2) = 4$, $\mathcal{V}(\psi_2^\perp) = P \cup L$, where P is a reduced point and L is a line not passing through P and (ψ_3^\perp) defines the unique minimal apolar set.



ψ has three essential variables ($h_\psi(1) = 3$) and a minimal apolar set Ξ of type (c):



- (i) if $d = 3$, then $\mathcal{V}(\psi_2^\perp) = \emptyset$ and \mathcal{W}_ψ is dense in the plane of essential variables;
- (ii) if $d \geq 4$, there is a unique minimal apolar set of points given by $I(\Xi) = (\psi_2^\perp)$.



ψ has four essential variables:

there is a unique minimal apolar set of points given by $I(\Xi) = (\psi_2^\perp)$.

Classification/algorithm for rank ≤ 5

HILBERT SEQUENCE	EXTRA CONDITION	ALGORITHM TO FIND A MINIMAL APOLAR SET
(1) [1*]		$\text{rk}(f) = 1$ and (f_1^\pm) defines the point apolar to f
(2) [1, 2, *, 2, 1]		f has two essential variables and Sylvester algorithm is applied: (i) if $f_{(f)}$ defines a set of 1(f) reduced points, then $\text{rk}(f) = 1(f)$; (ii) otherwise, $\text{rk}(f) = d + 2 - 1(f)$ and a minimal apolar set is given by the principal ideal generated by a generic form $g \in f_{d+2-1(f)}$
(3) [1, 3, 3, 1]	$Z(f_2^\pm) = \emptyset$ $Z(f_2^\pm) = P \cup D$, P is simple point D connected, 0-dim $\deg(D) = 2$	a generic pair of conics q_1, q_2 of f_2^\pm defines 4 points and $\text{rk}(f) = 4$ $\text{rk}(f) = 4$ and P is a point of any minimal apolar set; then, we find the scalar c such that $f' = f - ct_p^3$ has two essential variables and we apply Sylvester algorithm to f' as in (2)
(4) [1, 3, 3, 1]	$Z(f_2^\pm) = D$ D connected, 0-dim $\deg(D) = 2$	$\text{rk}(f) = 5$ and, for a generic P and a generic $c \neq 0$ such that $f' = f + ct_p^3$ is a ternary cubic of rank 4 and we apply (4) to f'
(5) [1, 3, 3, 1]	$Z(f_2^\pm) = \{P_1, P_2, P_3\}$ P_i 's are simple points $Z(f_2^\pm) = P \cup L$ P is simple point L is line, $P \notin L$	$\text{rk}(f) = 3$ and the unique minimal apolar set is $Z(f_2^\pm)$ P is a point of any minimal apolar set; then, we find the scalar c such that $f' = f - ct_p^4$ has two essential variables and we apply Sylvester algorithm to f' as in (2)
(6) [1, 3, 3*, 3, 1]	$Z(f_2^\pm) = \{P_1, \dots, P_k\}$ P_i 's are simple points	$\text{rk}(f) = 4$ and the unique minimal apolar set is $Z(f_2^\pm)$
(7) [1, 3, *, 3, 1]	$Z(f_2^\pm) = C$ C is irreducible quadric	let P be a generic point on C and c be a scalar such that $f' = f - ct_P^4$ has $h_{f'}(2) = 4$. (i) if $Z((f')_2^\pm) = \{P_1, \dots, P_4\}$ is a set of 4 reduced points, then, $\text{rk}(f) = 5$, and a minimal set apolar to f is $\{P, P_1, \dots, P_4\}$; (ii) otherwise, $\text{rk}(f) > 5$
(8) [1, 3, 4*, 3, 1]	$Z(f_2^\pm) = L_1 \cup L_2$ L_i 's are distinct lines	let P_i be a generic point on L_i , for $i = 1, 2$, respectively, and c_i be a scalar such that $f_i = f - c_i t_{P_i}^4$ has $h_{f_i}(2) = 4$, for $i = 1, 2$. (i) if $Z((f^i)_2) = \{P_1, \dots, P_4\}$, for either $i = 1$ or $i = 2$, then, $\text{rk}(f) = 5$, and a minimal apolar set of f is $\{P, P_1, \dots, P_4\}$; (ii) otherwise, $\text{rk}(f) > 5$
(9) [1, 3, 5, 3, 1]	$Z(f_2^\pm) = \{P_1, \dots, P_5\}$ P_i 's are reduced points	$\text{rk}(f) = 5$ and the unique minimal apolar set is $Z(f_2^\pm)$
(10) [1, 4, 4, 1]	$Z(f_2^\pm) = P \cup H$ P is a reduced point H is a plane, $P \notin H$	P is a point of any minimal apolar set; then, we find the scalar c such that $f' = f - ct_p^5$ has three essential variables and we apply (3) or (4) to f'
(11) [1, 4, 5*, 4, 1]	$Z(f_2^\pm) = \{P_1, \dots, P_5\}$	$\text{rk}(f) = 5$ and the unique minimal apolar set is $Z(f_2^\pm)$
(12) [1, 5, 5*, 5, 1]	$Z(f_2^\pm) = \{P_1, \dots, P_5\}$	$\text{rk}(f) = 5$ and the unique minimal apolar set is $Z(f_2^\pm)$

High rank and small forbidden locus

Definition: generic rank = rank of tensors on a dense open subset of the set of tensors.

Theorem (Alexander, Hirschowitz, 1995)

The generic rank of a tensor in $\mathbb{K}[x_0, \dots, x_n]_d$ is $\lceil \frac{1}{n+1} \binom{n+d}{d} \rceil$, except for $d = 2$ and $(n, d) \in \{(2, 4), (3, 4), (4, 3), (4, 4)\}$.

Theorem (Oneto, ., 2018)

Let g be the generic rank of tensors of degree d in \mathbb{P}^n . Let $\psi \in S_d$ with $r = \text{rank}(\psi)$. If $r > g$, then \mathcal{W}_ψ is dense in \mathbb{P}^n .

- ① Decomposition algorithms
- ② Low rank classification
- ③ **The variety of missing moments**

Flat extension of a truncated moment matrix

For (monomial) sets $B \subset C$, $B' \subset C'$, $\bar{B} = C \setminus B$, $\bar{B}' = C' \setminus B'$.

$$\mathbf{H}_\sigma^{\mathbf{C}, \mathbf{C}'} = \left(\langle \sigma \mid \mathbf{x}^{\alpha+\beta} \rangle \right)_{\alpha \in C, \beta \in C'} = \left[\begin{array}{c|c} \mathbf{H}_\sigma^{\mathbf{B}, \mathbf{B}'} & \mathbf{H}_\sigma^{\mathbf{B}, \bar{\mathbf{B}}'} \\ \hline \mathbf{H}_\sigma^{\bar{\mathbf{B}}, \mathbf{B}'} & \mathbf{H}_\sigma^{\bar{\mathbf{B}}, \bar{\mathbf{B}}'} \end{array} \right],$$

when

$$\text{rank } \mathbf{H}_\sigma^{\mathbf{C}, \mathbf{C}'} = \text{rank } \mathbf{H}_\sigma^{\mathbf{B}, \mathbf{B}'}$$

For $B \subset \mathbb{K}[\mathbf{x}]$, let $B^+ = B \cup x_1 B \cdots x_n B$, $\partial B = B^+ \setminus B$.

Theorem

Assume $H_\sigma^{B, B'}$ invertible with $|B| = |B'| = r$ and $C \supset B^+$, $C' \supset B'^+$ connected to 1 ($m \in C \Rightarrow m = 1$ or $m = x_j m'$ with $m' \in C'$).

$H_\sigma^{C, C'}$ is a **flat extension** of $H_\sigma^{B, B'}$

\Leftrightarrow The operators $M_j := H_\sigma^{B, x_j B} (H_\sigma^{B, B})^{-1}$ **commute**.

$\Leftrightarrow \exists! \tilde{\sigma} \in \mathcal{P}\text{olExp}$ s.t. $\text{rank } H_{\tilde{\sigma}} = r$ and $\tilde{\sigma}|_{C, C'} = \sigma$.

Example

$$\begin{aligned} \sigma &= 8 + 17 z_2 - 4 z_1 + 15 z_2^2 + 14 z_1 z_2 - 16 z_1^2 + 47 z_2^3 - 6 z_1 z_2^2 \\ &\quad + 34 z_1^2 z_2 - 52 z_1^3 + 51 z_2^4 + 38 z_1 z_2^3 - 18 z_1^2 z_2^2 + 86 z_1^3 z_2 - 160 z_1^4 \end{aligned}$$

moment series $\in \mathbb{K}[[z_1, z_2]]$, truncated in degree 4.

$$[H_\sigma^{B^+, B^+}] = \begin{bmatrix} 8 & -4 & 17 & -16 & 14 & 15 & -52 & 34 & -6 & 47 \\ -4 & -16 & 14 & -52 & 34 & -6 & -160 & 86 & -18 & 38 \\ 17 & 14 & 15 & 34 & -6 & 47 & 86 & -18 & 38 & 51 \\ -16 & -52 & 34 & -160 & 86 & -18 & h_1 & h_2 & h_3 & h_4 \\ 14 & 34 & -6 & 86 & -18 & 38 & h_2 & h_3 & h_4 & h_5 \\ 15 & -6 & 47 & -18 & 38 & 51 & h_3 & h_4 & h_5 & h_6 \\ -52 & -160 & 86 & h_1 & h_2 & h_3 & h_7 & h_8 & h_9 & h_{10} \\ 34 & 86 & -18 & h_2 & h_3 & h_4 & h_8 & h_9 & h_{10} & h_{11} \\ -6 & -18 & 38 & h_3 & h_4 & h_5 & h_9 & h_{10} & h_{11} & h_{12} \\ 47 & 38 & 51 & h_4 & h_5 & h_6 & h_{10} & h_{11} & h_{12} & h_{13} \end{bmatrix}$$

where $B = \{1, x_1, x_2, x_1^2, x_1 x_2, x_2^2\}$.

Flat extension condition: $\text{rank} H_{\sigma}^{B^+, B^+} \leq 6$ implies

$$\left\{ \begin{array}{l} -814592 h_1^2 - 1351680 h_1 h_2 - 476864 h_1 h_3 - 599040 h_2^2 - 301440 h_2 h_3 - 35072 h_3^2 \\ \quad - 520892032 h_1 - 396821760 h_2 - 164529152 h_3 + 1693440 h_7 - 86394672128 = 0 \\ -814592 h_2^2 - 1351680 h_2 h_3 - 476864 h_2 h_4 - 599040 h_3^2 - 301440 h_3 h_4 - 35072 h_4^2 \\ \quad + 335275392 h_2 + 257276160 h_3 + 96277632 h_4 + 1693440 h_9 - 34904464128 = 0 \\ \vdots \\ -814592 h_1 h_3 - 675840 h_1 h_4 - 238432 h_1 h_5 - 675840 h_2 h_3 - 599040 h_2 h_4 - 150720 h_2 h_5 - 238432 h_3^2 \\ \quad - 150720 h_3 h_4 - 35072 h_3 h_5 + 6613440 h_1 + 6641280 h_2 - 264559616 h_3 - 198410880 h_4 - 82264576 h_5 \\ \quad + 1693440 h_9 + 1312368000 = 0 \\ -814592 h_1 h_4 - 675840 h_1 h_5 - 238432 h_1 h_6 - 675840 h_2 h_4 - 599040 h_2 h_5 - 150720 h_2 h_6 - 238432 h_3 h_4 \\ \quad - 150720 h_3 h_5 - 35072 h_3 h_6 + 106430368 h_1 + 81349440 h_2 + 25713728 h_3 - 260446016 h_4 \\ \quad - 198410880 h_5 - 82264576 h_6 + 1693440 h_{10} + 34550702464 = 0 \end{array} \right.$$

Solution set: an algebraic variety of dimension 3 and degree 52.

A solution (among others) is $h_1 = -484$, $h_2 = 226$, $h_3 = -54$, $h_4 = 82$, $h_5 = -6$, $h_6 = 167$,

$h_7 = -1456$, $h_8 = 614$, $h_9 = -162$, $h_{10} = 182$, $h_{11} = -18$, $h_{12} = 134$, $h_{13} = 195$.

Decomposition of **rank 6** of the series with these computed moments:

$$\begin{aligned} \sigma &\equiv (0.517 + 0.044 i) \mathbf{e}_{-0.830+1.593 i, -0.326-0.050 i} + (0.517 - 0.044 i) \mathbf{e}_{-0.830-1.593 i, -0.326+0.050 i} \\ &\quad + 2.958 \mathbf{e}_{1.142, 0.836} + 4.583 \mathbf{e}_{0.956, -0.713} \\ &\quad - (4.288 + 1.119 i) \mathbf{e}_{-0.838+0.130 i, 0.060+0.736 i} - (4.288 - 1.119 i) \mathbf{e}_{-0.838-0.130 i, 0.060-0.736 i} \end{aligned}$$

General decomposition algorithm [BCMT10], [BS18]

- ▶ Perform a generic change of coordinates $\psi'(\bar{x}) = \psi(T \bar{x})$.
- ▶ For $r = \max \text{rank} H_\sigma^{k, d-k}, \dots$
 - ▶ For bases B, B' of size r , connected to 1 (e.g. B stable by division/Borel fixed stable by division);
 - ① **Find the (unknown) moments** of $H_\Lambda^{B'+, B+}$ s.t.
 - $H_\Lambda^{B', B}$ invertible and
 - the operators $M_i = H_\Lambda^{x_i B', B} (H_\Lambda^{B', B})^{-1}$ commute.
 - ② Deduce the decomposition of σ (Algorithm 1).
 - ③ If the roots are simple and the decomposition is valid for the moments of ψ , **stop and output a decomposition of ψ** ;

Challenges, open questions

- ▶ Numerical stability, correction of errors,
- ▶ Efficient construction of basis, complexity,
- ▶ Super-resolution, collision of points,
- ▶ Super-extrapolation,
- ▶ Best low rank approximation,
- ▶ ...

Thanks for your attention

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