### Trivariate polynomial approximation on Lissajous curves <sup>1</sup>

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<sup>1</sup>Joint work with Len Bos (Verona) and Marco Vianello (Padova)



### 1 Introduction

- 2 Looking for "good" 3d Lissajous curves
- 3 Hyperinterpolation on Lissajous curves
- 4 Computational issues
- 5 Interpolation

#### 6 To do

### Lissajous curves

Properties and motivation



Are parametric curves studied by Bowditch (1815) and Lissajous (1857) of the form

$$\gamma(t) = (A\cos(at + \alpha), B\sin(bt + \beta)).$$

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$$\begin{cases} y = -\sin\left(nt - \frac{\pi}{2}\right) & 0 \le t \le \pi \end{cases}$$

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- **3** Padua points (of the first family) lie on  $[-1, 1]^2$  on the  $\pi$ -periodic Lissajous curve  $T_{n+1}(x) = T_n(y)$  called generating curve given also as

$$\gamma_n(t) = (\cos nt, \cos(n+1)t), \quad 0 \le t \le \pi, \ n \ge 1.$$

# The generating curve of the Padua points (n = 4)

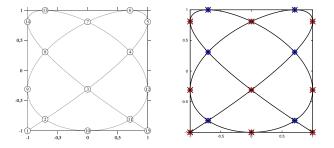


Figure :  $\operatorname{Pad}_{n} = C_{n+1}^{O} \times C_{n+2}^{E} \cup C_{n+1}^{E} \times C_{n+2}^{O} \subset C_{n+1} \times C_{n+2}$ 

 $C_{n+1} = \{z_j^n = \cos\left(\frac{(j-1)\pi}{n}\right), j = 1, \dots, n+1\}$ : Chebsyhev-Lobatto points on [-1, 1]

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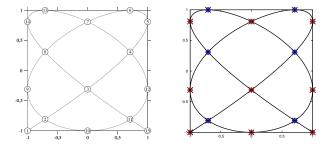


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$$\begin{split} & C_{n+1} = \left\{ z_j^n = \cos\left(\frac{(j-1)\pi}{n}\right), \ j = 1, \dots, n+1 \right\}: \text{Chebsyhev-Lobatto points} \\ & \text{on } [-1,1] \\ & \text{Note: } |Pad_n| = \binom{n+2}{2} = dim(\mathbb{P}_n(\mathbb{R}^2)) \end{split}$$



### Lemma (cf. Bos at al. JAT 2006)

For all  $p \in \mathbb{P}_{2n}(\mathbb{R}^2)$  we have

$$\frac{1}{\pi^2}\int_{[-1,1]^2}p(x,y)\frac{1}{\sqrt{1-x^2}}\frac{1}{\sqrt{1-y^2}}dxdy=\frac{1}{\pi}\int_0^{\pi}p(\gamma_n(t))dt.$$



Erb et al. 2014 (cf. arXiv 1411.7589) in the framework of Magnetic Particle Imaging applications, considered

$$\gamma_{n,p}(t) = (\sin nt, \sin((n+p)t)) \quad 0 \le t < 2\pi,$$

 $n, p \in \mathbb{N}$  s.t. *n* and n + p are relative primes.

•  $\gamma_{n,p}$  is non-degenerate iff p is odd.

Take 
$$t_k = 2\pi k/(4n(n+p)), k = 1, ..., 4n(n+p).$$

 $Lisa_{n,p} := \{\gamma_{n,p}(t_k), k = 1, ..., 4n(n+p)\}, |Lisa_{n,p}| = 2n(n+p)+2n+p.$ 

Notice: p = 1,  $|Lisa_{n,1}| = 2n^2 + 4n + 1$  while  $|Pad_{2n}| = 2n^2 + 3n + 1$ .

### Lissajous points: non-degenerate case

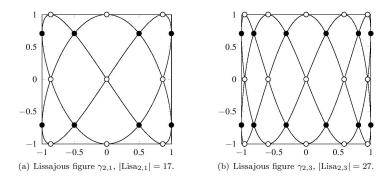


Figure : From the paper by Erb et al. 2014 (cf. arXiv 1411.7589)

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### Lissajous points: degenerate case



### Erb 2015, (cf. arXiv 1503.00895) has then studied the degenerate $2\pi$ -Lissajous curves

$$\gamma_{n,p}(t) = (\cos nt, \cos((n+p)t)) \quad 0 \le t < 2\pi,$$

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Take 
$$t_k = \pi k / (n(n+p)), \ k = 0, 1, ..., n(n+p).$$

$$LD_{n,p} := \{\gamma_{n,p}(t_k), k = 0, 1, ..., n(n+p)\}, |LD_{n,p}| = \frac{(n+p+1)(n+1)}{2}$$

Notice: for p = 1,  $|LD_{n,1}| = |Pad_n| = \dim(\mathbb{P}_n(\mathbb{R}^2))$  and correspond to the Padua points themselves.

### Lissajous points: degenerate case



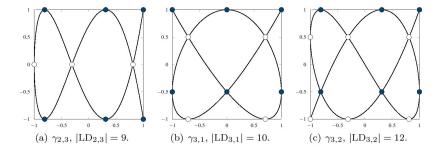


Figure : From the paper by Erb 2015, (cf. arXiv 1503.00895)



•  $\Omega = [-1, 1]^3$ : the standard 3-cube

The product Chebyshev measure

$$d\lambda = w(\mathbf{x})d\mathbf{x}, \quad w(\mathbf{x}) = \frac{1}{\sqrt{(1-x_1^2)(1-x_2^2)(1-x_3^2)}}.$$
 (1)

■  $\mathbb{P}_k^3$ : space of trivariate polynomials of degree k in  $\mathbb{R}^3$ (dim( $\mathbb{P}_k^3$ ) = (k + 1)(k + 2)(k + 3)/6).





This results shows which are the admissible 3d Lissajous curves

#### Theorem (cf. Bos, De Marchi at al. 2015, arXiv 1502.04114)

Let be  $n \in \mathbb{N}^+$  and  $(a_n, b_n, c_n)$  be the integer triple

$$(a_n, b_n, c_n) = \begin{cases} \left(\frac{3}{4}n^2 + \frac{1}{2}n, \frac{3}{4}n^2 + n, \frac{3}{4}n^2 + \frac{3}{2}n + 1\right), & n \text{ even} \\ \left(\frac{3}{4}n^2 + \frac{1}{4}, \frac{3}{4}n^2 + \frac{3}{2}n - \frac{1}{4}, \frac{3}{4}n^2 + \frac{3}{2}n + \frac{3}{4}\right), & n \text{ odd} \end{cases}$$
(2)

Then, for every integer triple (i, j, k), not all 0, with  $i, j, k \ge 0$  and  $i + j + k \le m_n = 2n$ , we have the property that  $ia_n \ne jb_n + kc_n$ ,  $jb_n \ne ia_n + kc_n$ ,  $kc_n \ne ia_n + jb_n$ . Moreover,  $m_n = 2n$  is maximal, in the sense that there exists a triple  $(i^*, j^*, k^*)$ ,  $i^* + j^* + k^* = 2n + 1$ , that does not satisfy the property.



Cubature along the curve



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On admissible curve the integration becomes the integral on the curve.

### Proposition

Consider the Lissajous curves in  $[-1, 1]^3$  defined by

$$\ell_n(\theta) = (\cos(a_n\theta), \cos(b_n\theta), \cos(b_n\theta)), \ \theta \in [0, \pi], \quad (3)$$

where  $(a_n, b_n, c_n)$  is the sequence of integer triples (2). Then, for every total-degree polynomial  $p \in \mathbb{P}^3_{2n}$ 

$$\int_{[-1,1]^3} p(\boldsymbol{x}) w(\boldsymbol{x}) d\boldsymbol{x} = \pi^2 \int_0^{\pi} p(\boldsymbol{\ell}_n(\theta)) d\theta .$$
 (4)

**Proof.** It suffices to prove the identity for a polynomial basis (ex: for the tensor product basis  $T_{\alpha}(\mathbf{x}), |\alpha| \leq 2n$ ).  $\Box$ 

### **Consequence II**

Exactness

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### Corollary

Let  $p \in \mathbb{P}^3_{2n}$ ,  $\ell_n(\theta)$  and  $\nu = n \max\{a_n, b_n, c_n\}$ . Then

$$\int_{[-1,1]^3} p(\boldsymbol{x}) w(\boldsymbol{x}) d\boldsymbol{x} = \sum_{s=0}^{\mu} w_s p(\boldsymbol{\ell}_n(\theta_s)) , \qquad (5)$$

#### where

$$w_s = \pi^2 \omega_s , \quad s = 0, \dots, \mu , \qquad (6)$$

with

$$\mu = \nu , \quad \theta_{s} = \frac{(2s+1)\pi}{2\mu + 2} , \quad \omega_{s} \equiv \frac{\pi}{\mu + 1} , \quad s = 0, \dots, \mu ,$$
 (7)

or alternatively

$$\mu = \nu + 1 , \quad \theta_{s} = \frac{s\pi}{\mu} , \quad s = 0, \dots, \mu ,$$
  
$$\omega_{0} = \omega_{\mu} = \frac{\pi}{2\mu} , \quad \omega_{s} \equiv \frac{\pi}{\mu} , \quad s = 1, \dots, \mu - 1 .$$
(8)



#### The points set

$$\{\boldsymbol{\ell}_n(\theta_s), \ s=0,\ldots,\mu\}$$

are a 3-dimensional rank-1 Chebyshev lattices (for cubature of degree of exactness 2*n*).

- Cools and Poppe [cf. CHEBINT, TOMS 2013] wrote a search algorithm for constructing heuristically such lattices.
- WE HAVE an explicit formula for any degree.

General definition



#### Definition

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It requires 3 main ingredients

- **1** a good cubature formula (positive weights and high precision);
- **2** a good formula for representing the reproducing kernel (accurate and efficient);
- **3** a slow increase of the Lebesgue constant (which is the operator norm).





For  $f \in C([-1, 1]^3)$ , using (5), the hyperinterpolation polynomial of f is

$$\mathcal{H}_n f(\mathbf{x}) = \sum_{0 \le i+j+k \le n} C_{i,j,k} \,\hat{\phi}_{i,j,k}(\mathbf{x}) , \qquad (9)$$

 $\hat{\phi}_{i,j,k}(\mathbf{x}) = \hat{T}_i(x_1)\hat{T}_j(x_2)\hat{T}_k(x_3)$  with  $T_s(\cdot)$  the normalized Chebyshev polynomials

$$C_{i,j,k} = \sum_{s=0}^{\mu} w_s f(\boldsymbol{\ell}_n(\theta_s)) \, \hat{\phi}_{i,j,k}(\boldsymbol{\ell}_n(\theta_s)) \; . \tag{10}$$

### Properties



*H<sub>n</sub>f* = *f*, ∀*f* ∈ P<sup>3</sup><sub>n</sub> (projection operator, by construction).
 *L*<sup>2</sup>-error

$$|f - \mathcal{H}_n f||_2 \le 2\pi^3 \, E_n(f) \,, \quad E_n(f) = \inf_{p \in \mathbb{P}_n} ||f - p||_{\infty} \,. \tag{11}$$

Lebesgue constant

$$\|\mathcal{H}_{n}\|_{\infty} = \max_{\boldsymbol{x} \in [-1,1]^{3}} \sum_{s=0}^{\mu} w_{s} \left| K_{n}(\boldsymbol{x}, \boldsymbol{\ell}_{n}(\boldsymbol{\theta}_{s})) \right|, \ K_{n}(\boldsymbol{x}, \boldsymbol{y}) = \sum_{|\mathbf{i}| \le n} \hat{\phi}_{\mathbf{i}}(\boldsymbol{x}) \hat{\phi}_{\mathbf{i}}(\boldsymbol{y}),$$
(12)
where  $K_{n}$  is the reproducing kernel of  $\mathbb{P}_{n}^{3}$  w.r.t. product

Chebyshev measure





Norm and approximation error estimates

 Based on a conjecture stated in [DeM, Vianello & Xu, BIT 2009] and specialized in [H.Wang, K.Wang & X.Wang, CMA 2014] we get

$$\|\mathcal{H}_n\|_{\infty} = O((\log n)^3)$$

i.e. the minimal polynomial growth.

 $\blacksquare$   $\mathcal{H}_n$  is a projection, then

$$\|f - \mathcal{H}_n f\|_{\infty} = O\left((\log n)^3 E_n(f)\right) . \tag{13}$$

### Computing the hyperinterpolation coefficient

The coefficients  $\{C_{i,j,k}\}$  can be computed by a single 1D discrete Chebyshev transform along the Lissajous curve.

### Proposition

Let be  $f \in C([-1, 1]^3)$ ,  $(a_n, b_n, c_n)$ ,  $v, \mu, \{\theta_s\}$ ,  $\omega_s$ ,  $\{w_s\}$  as in Corollary 1. Then

$$C_{i,j,k} = \frac{\pi^2}{4} \sigma_{ia_n} \sigma_{jb_n} \sigma_{kc_n} \left( \frac{\gamma_{\alpha_1}}{\sigma_{\alpha_1}} + \frac{\gamma_{\alpha_2}}{\sigma_{\alpha_2}} + \frac{\gamma_{\alpha_3}}{\sigma_{\alpha_3}} + \frac{\gamma_{\alpha_4}}{\sigma_{\alpha_4}} \right), \quad (14)$$

$$\alpha_1 = ia_n + jb_n + kc_n , \quad \alpha_2 = |ia_n + jb_n - kc_n| ,$$

$$\alpha_3 = |ia_n - jb_n| + kc_n$$
,  $\alpha_4 = ||ia_n - jb_n| - kc_n|$ ,

where { $\gamma_m$ } are the first v + 1 coefficients of the discretized Chebyshev expansion of  $f(T_{a_n}(t), T_{b_n}(t), T_{c_n}(t))$ ,  $t \in [-1, 1]$ , namely

$$\gamma_{m} = \sum_{s=0}^{\mu} \omega_{s} \, \hat{T}_{m}(\tau_{s}) \, f(T_{a_{n}}(\tau_{s}), T_{b_{n}}(\tau_{s}), T_{c_{n}}(\tau_{s})) \,, \tag{15}$$

 $m = 0, 1, ..., \nu$ , with  $\tau_s = \cos(\theta_s)$ ,  $s = 0, 1, ..., \mu$ .



- Within the emerging 3d MPI (Magnetic Particle Imaging) technology, Lissajous sampling is one of the most common sampling method.
- 2 Hyperinterpolation polynomials on *d*-dimensional cubes can be done by other cubature formulas (exacteness 2n) for the product Chebyshev measure :  $O(n^4)$  [DeM, Vianello, Xu BIT 2009]; [Godzina 1995] which have the lowest cardinality and used in the package CHEBINT by Cools&Poppe.
- **3** The set  $\mathcal{A}_n = \{\ell_n(\theta_s), s = 0, ..., \mu\}$  in (7)-(8), forms a Weakly Admissible Mesh for  $[-1, 1]^3$ , so they provide a Lissajous sampling to 3D polynomial interpolation.

From Prop. 2, hyperinterpolation on  $\ell_n(t)$  can be done by a single 1-dimensional FFT. We used the Chebfun package [Chebfun 2014].

The polynomial interpolant of a function g can be written

$$\pi_{\mu}(t) = \sum_{m=0}^{\mu} c_m T_m(t)$$
 (16)

where

$$c_{m} = \frac{2}{\mu} \sum_{s=0}^{\mu} {}^{\prime\prime} T_{m}(\tau_{s}) g(\tau_{s}) , \quad m = 1, \dots, \mu - 1 ,$$
  
$$c_{m} = \frac{1}{\mu} \sum_{s=0}^{\mu} {}^{\prime\prime} T_{m}(\tau_{s}) g(\tau_{s}) , \quad m = 0, \mu , \qquad (17)$$

**Note**:  $\sum_{s=0}^{\mu}$  " means first and last terms are halved





If g(t) = f(T<sub>a<sub>n</sub></sub>(t), T<sub>b<sub>n</sub></sub>(t), T<sub>c<sub>n</sub></sub>(t)) and comparing with the discrete Chebyshev expansion coefficients (15)

$$\frac{\gamma_m}{\sigma_m} = \begin{cases} \frac{\pi}{2} c_m, & m = 1, \dots, \mu - 1\\ \pi c_m, & m = 0, \mu \end{cases}$$
(18)

i.e., the 3D hyperinterpolation coefficients (14) can be computed by the  $\{c_m\}$  and (18).

• A single call of the function **chebfun** on  $f(T_{a_n}(t), T_{b_n}(t), T_{c_n}(t))$ , truncated at the  $(\mu + 1)$ th-term, produces all the relevant coefficients  $\{c_m\}$  in an extremely fast and stable way.



#### Example

Take n = 100 and the functions

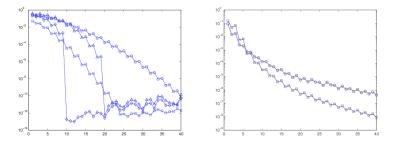
$$f_1(\mathbf{x}) = \exp(-c \|\mathbf{x}\|_2^2), \ c > 0, \ f_2(\mathbf{x}) = \|\mathbf{x}\|_2^{\beta}, \ \beta > 0,$$
 (19)

To compute the  $\mu = \frac{3}{4}n^3 + \frac{3}{2}n^2 + n + 2 = 765102$  coefficients from which we get by (14) the (n + 1)(n + 2)(n + 3)/6 = 176851 coefficients of trivariate hyperinterpolation,

it took about 1 second by using Chebfun 5.1 on a Athlon 64 X2 Dual Core 4400+ 2.4 GHz processor.

### Example hyperinterpolation errors





**Figure** : Left: Hyperinterpolation errors for the trivariate polynomials  $\|\mathbf{x}\|_2^{2k}$  with k = 5 (diamonds) and k = 10 (triangles), and for the trivariate function  $f_1$  with c = 1 (squares) and c = 5 (circles). Right: Hyperinterpolation errors for the trivariate function  $f_2$  with  $\beta = 5$  (squares) and  $\beta = 3$  (circles).



- The sampling set (Chebyshev lattice)  $\mathcal{A}_n = \{\ell_n(\theta_s), \ s = 0, ..., \mu\} \text{ has been used as a WAM from which we have extracted the AFP and the DLP.}$
- The extraction of N = dim(P<sup>3</sup><sub>n</sub>) points has been done by the software available at www.math.unipd.it/~marcov/CAAsoft.
   DLP form a sequence, i.e., its first N<sub>r</sub> = dim(P<sup>d</sup><sub>r</sub>) elements span P<sup>d</sup><sub>r</sub>, 1 ≤ r ≤ n.
- We wrote the package hyperlissa, a Matlab code for hyperinterpolation on 3d Lissajous curves. Available at the same web page.

### Example Chebyshev lattice points



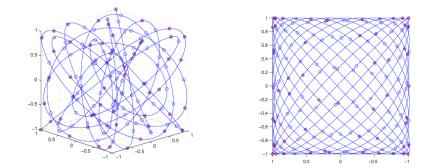


Figure : Left: the Chebyshev lattice (circles) and the extracted Approximate Fekete Points (red asterisks), on the Lissajous curve for polynomial degree n = 5. Right: A face projection of the curve and the sampling nodes

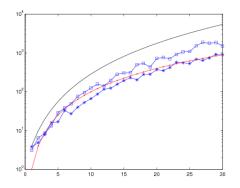


Figure : Lebesgue constants (log scale) of the Approximate Fekete Points (asterisks) and Discrete Leja Points (squares) extracted from the Chebyshev lattices on the Lissajous curves, for degree n = 1, 2, ..., 30, compared with dim $(\mathbb{P}_n^3) = (n+1)(n+2)(n+3)/6$  (upper solid line) and  $n^2$  (dots).

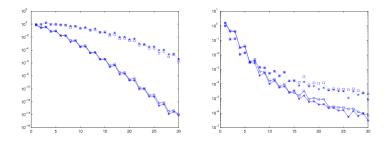


Figure : Interpolation errors on Approximate Fekete Points (asterisks) and Discrete Leja Points (squares) for the trivariate functions  $f_1$  (Left) with c = 1 (solid line) and c = 5 (dotted line), and  $f_2$  (Right) with  $\beta = 5$  (solid line) and  $\beta = 3$  (dotted line).



- A general results for generating "good" Lissajous curves for every dimension *d*.
- A faster way to extract AFP and LDP from Lissajous curves.
- Construct Padua points on the cube  $[-1, 1]^3$ .



## 7th Research week: 5-8/9/2015, http://events.math.unipd.it/drwa15/

### 4th Workshop: 8-13/9/2016, http://events.math.unipd.it/dwcaa16/

#thankyou!