

# Computing the common zeros of two bivariate functions via Bézout resultants

Colorado State University, 26th September 2013

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# Introduction

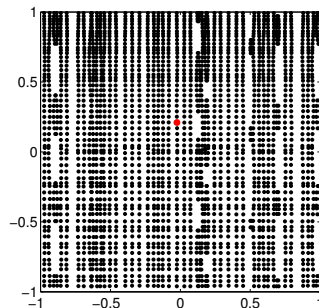
## Motivation

- Global 1D rootfinding is crucial (25% of Chebfun code needs roots)
- Chebfun2 is an extension of Chebfun for bivariate functions
- Very high degree polynomial interpolants are common

Find the global minimum of

$$\begin{aligned} f(x, y) = & \left( \frac{x^2}{4} + e^{\sin(50x)} + \sin(70 \sin(x)) \right) \\ & + \left( \frac{y^2}{4} + \sin(60e^y) + \sin(\sin(80y)) \right) \\ & - \cos(10x) \sin(10y) - \sin(10x) \cos(10y). \end{aligned}$$

```
g = chebfun2( f );  
r = roots( gradient( g ) );
```



There are 2720 local extrema.

# Introduction

## Algorithmic overview

Let  $f$  and  $g$  be real-valued Lipschitz functions on  $[-1, 1]^2$ . Solve:

$$\begin{pmatrix} f(x, y) \\ g(x, y) \end{pmatrix} = 0, \quad (x, y) \in [-1, 1]^2.$$

- **“Polynomialization”**: Replace  $f$  and  $g$  with bivariate polynomials  $p$  and  $q$
- **“Act locally”**: Subdivide  $[-1, 1]^2$  with piecewise approximants until total degree  $\leq 16$ , solve low degree rootfinding problems
- **“Think globally”**: Do refinement and regularization to improve global stability

**“Think globally, act locally”**, Stan Wagon



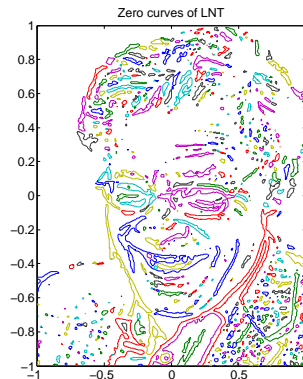
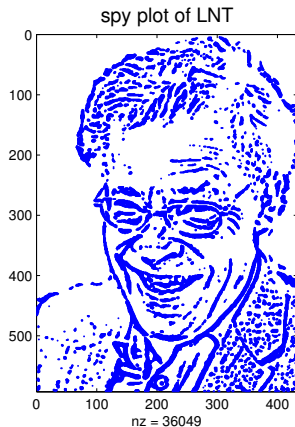
# Introduction

## NOT curve finding

Not to be confused with bivariate ~~root finding~~ curve finding:

$$f(x, y) = 0, \quad (x, y) \in [-1, 1]^2.$$

Solutions lie along curves. Chebfun2 computes these by **Marching Squares**.



\* Photo courtesy of Nick Hale.

# Introduction

## Talk overview

The talk follows Stan Wagon:

- “Polynomialization”
- “Act locally”
- “Think globally”
- Numerical examples



**WARNING:** Simple common zeros only!

# Polynomialization

## 1D Chebyshev interpolants

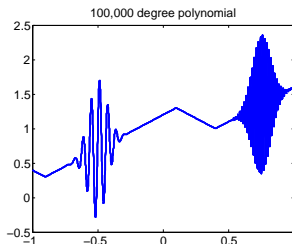
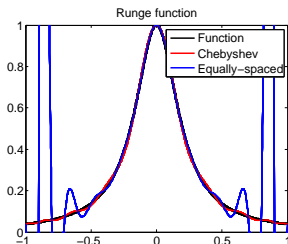
For  $n \geq 1$ , the **Chebyshev points** (of the 2nd kind) are given by

$$x_j^n = \cos\left(\frac{j\pi}{n}\right), \quad 0 \leq j \leq n.$$

The **Chebyshev interpolant** of  $f$  is the polynomial  $p$  of degree at most  $n$  s.t.

$$p(x) = \sum_{j=0}^n c_j T_j(x), \quad p(x_j^n) = f(x_j^n), \quad 0 \leq j \leq n,$$

where  $T_j(x) = \cos(j \cos^{-1}(x))$  is the Chebyshev polynomial of degree  $j$ .



# Polynomialization

## Tensor-product approximation

Replace  $f$  and  $g$  by their polynomial interpolants

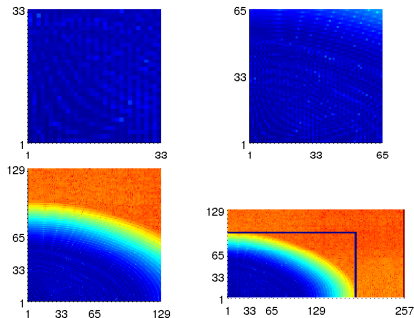
$$p(x, y) = \sum_{i=0}^{n_p} \sum_{j=0}^{m_p} \alpha_{ij} T_i(x) T_j(y), \quad q(x, y) = \sum_{i=0}^{n_q} \sum_{j=0}^{m_q} \beta_{ij} T_i(x) T_j(y)$$

such that  $p(x_s^{n_p}, x_t^{m_p}) = f(x_s^{n_p}, x_t^{m_p})$  and  $q(x_s^{n_q}, x_t^{m_q}) = g(x_s^{n_q}, x_t^{m_q})$ . Select  $n_p$ ,  $m_p$  and  $n_q$ ,  $m_q$  large enough.

Take  $n_p = 9, 17, 33, 65$ , and so on, until tail of coefficients falls below **relative** machine precision.

Chebyshev coefficients computed by fast DCT-I transform [Gentleman 72].

Chebyshev coefficients: Have we converged?



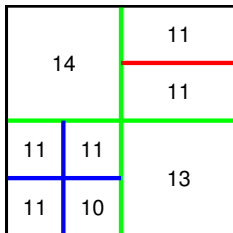
# Act locally

## Subdivision

**Key fact: Subdivide to deal with high degree**

Subdivide into subrectangles until polynomial degrees are small.

$$\begin{aligned}\sin((x-1/10)y)\cos(1/(x + (y-9/10) + 5)) \\ = (y-1/10)\cos((x+(y+9/10)^2/4)) = 0\end{aligned}$$



Real solutions only.

Do not bisect! Instead subdivide off-center (to avoid awkward coincidences).

Subdivide until degree 16.

Like 1D subdivision:





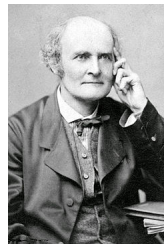
### Theorem (Bézout resultant theorem)

*Let  $p_y$  and  $q_y$  be two univariate polynomials of degree at most  $n_p$  and  $n_q$ . The Chebyshev Bézout resultant matrix*

$$B(p_y, q_y) = (b_{ij})_{1 \leq i, j \leq \max(n_p, n_q)}, \quad \frac{p_y(s)q_y(t) - p_y(t)q_y(s)}{s - t} = \sum_{i, j=1}^{\max(n_p, n_q)} b_{ij} T_{i-1}(s) T_{j-1}(t).$$

*is nonsingular if and only if  $p_y$  and  $q_y$  have no common roots.*

- Usually, this theorem is stated using the Sylvester resultant
- Usually, stated in terms of the monomial basis
- There are stable ways to form  $B(p_y, q_y)$ . We use [T., Noferini, Nakatsukasa, 13a]



# Act locally

## Hidden-variable resultant method

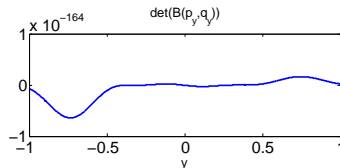
The hidden-variable resultant method “hides” one of the variables:

$$p_y(x) = p(x, y) = \sum_{i=0}^{n_p} \alpha_i(y) T_i(x), \quad q_y(x) = q(x, y) = \sum_{i=0}^{n_q} \beta_i(y) T_i(x).$$

- $B(p_y, q_y)$  is a **symmetric** matrix of size  $\max(n_p, n_q)$
- Each entry of  $B(p_y, q_y)$  is a polynomial in  $y$ , of degree  $m_p + m_q$
- For the  $y$ -values of  $p(x, y) = q(x, y) = 0$  we want to solve

$$\det(B(p_y, q_y)) = 0, \quad y \in [-1, 1].$$

**Problem!** Determinant is numerically zero:



# Act locally

## Matrix polynomial linearization

**Key fact: Inherit robustness from eigenvalue solver**

$B(p_y, q_y)$  is a matrix-valued polynomial in  $y$ :  $B(p_y, q_y) = \sum_{i=0}^M A_i T_i(y) \in \mathbb{R}^{N \times N}$ .

The colleague matrix [Specht 1960, Good 1961]:

$$yX + Y = y \begin{bmatrix} A_M & & & & \\ & I_N & & & \\ & & \ddots & & \\ & & & I_N & \end{bmatrix} - \frac{1}{2} \begin{bmatrix} -A_{M-1} & I_N - A_{M-2} & -A_{M-3} & \cdots & -A_0 \\ I_N & 0 & I_N & & \\ & \ddots & \ddots & \ddots & \\ & & I_N & 0 & I_N \\ & & & 2I_N & 0 \end{bmatrix}.$$



- Similar to companion, but for Chebyshev.
- **Inherited robustness** from eigenvalue solver.
- Strong linearization.



# Act locally

## Univariate rootfinding

**Key point: Use univariate rootfinder for  $x$ -values**

We use Chebfun's 1D rootfinder for the  $x$ -values, once we have the  $y$ -values.  
We independently solve for each  $y_*$

$$p(x, y_*) = 0, \quad x \in [-1, 1] \quad \text{and} \quad q(x, y_*) = 0, \quad x \in [-1, 1].$$



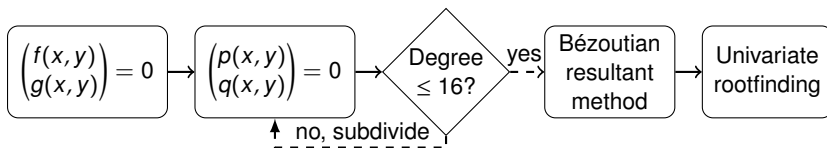
- Based on the colleague matrix ( $\approx$  companion matrix)
- Gets its robustness from eigenvalue solver
- Originally Boyd's algorithm from [Boyd 02]
- 1D subdivision is not needed for us



# Act locally

## Reviewing the algorithm

### Flowchart of the algorithm:



Collect together the solutions from the subdomains.

Keep solutions in  $[-1, 1]^2$ , throw away the rest. Perturb some if necessary.

### Further questions:

1. Should we hide the  $x$ - or  $y$ -variable in the hidden-variable resultant method?
2. What is the operational cost of the algorithm?
3. Is the algorithm stable?

# Think globally

## Stability of the Bézout resultant method

Let  $p(x_*, y_*) = q(x_*, y_*) = 0$  with  $\|p\|_\infty = \|q\|_\infty = 1$ . The Jacobian matrix is

$$J = J(x_*, y_*) = \begin{bmatrix} \frac{\partial p}{\partial x}(x_*, y_*) & \frac{\partial p}{\partial y}(x_*, y_*) \\ \frac{\partial q}{\partial x}(x_*, y_*) & \frac{\partial q}{\partial y}(x_*, y_*) \end{bmatrix}.$$

Absolute condition number of problem at  $(x_*, y_*)$ :  $\kappa_* = \|J^{-1}\|_2$

Absolute condition number of  $y_*$  for Bézout:  $\kappa(y_*, B) \geq \frac{1}{2} \frac{\kappa_*^2}{\kappa_2(J)} \geq \frac{\kappa_*}{\|\text{adj}(J)\|_2}$  [1]

**The Bézout resultant method is unstable:** If entries of  $J$  are small then,

$$\kappa(y_*, B) \gg \kappa_*$$

This is BAD news!

[1] Nakatsukasa, Noferini, & T., 2013b.

# Think globally

## Local refinement

**Key fact: Local refinement can improve stability**

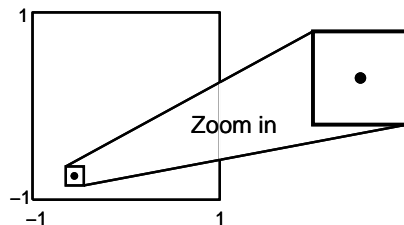
Redo Bézout resultant in  $\Omega$  near  $(x_*, y_*)$ .

Let  $\Omega = [x_{\min}, x_{\max}] \times [y_{\min}, y_{\max}]$

$$|\Omega| = x_{\max} - x_{\min} \approx y_{\max} - y_{\min}$$

$$\kappa_{\Omega}(y_*, B) \approx |\Omega|^2 \kappa(y_*, B)$$

- Shrinking  $|\Omega|$  improves stability (in a think globally sense).
- Get  $O(\kappa_* u)$  error from polynomialization.
- Also do local refinement in detected ill-conditioned regions.



**Key fact: Regularize the problem by projecting**

The Bézout resultant is symmetric. Partition such that

$$B(p_y, q_y) = \begin{bmatrix} B_1(y) & E(y)^T \\ E(y) & B_0(y) \end{bmatrix}, \quad B_1(y) \in \mathbb{R}^{k \times k}, \quad B_0(y) \in \mathbb{R}^{(N-k) \times (N-k)},$$

with

$$\|B_0(y)\|_2 = O(u), \quad \|E(y)\|_2 = O(u^{1/2}).$$

The eigenvalues of  $B_1(y)$  and  $B(p_y, q_y)$  in  $[-1, 1]$  are usually within  $O(u)$ .

Effectively this step removes large eigenvalues.



# More details

Many other approaches

## Homotopy continuation method

Solve a problem, make it harder.

$$H(\lambda, z) + Q(z)(1 - \lambda) + P(z)\lambda,$$
$$\lambda \in (0, 1).$$

## Contour algorithms

Solve two curve finding problems:

$$f(x, y) = 0, \quad g(x, y) = 0.$$

Find intersection of curves.

## Two-parameter eigenvalue problem

Use EIG to solve  $x$  and  $y$  together.

$$A_1 v = xB_1 v + yC_1 v,$$

$$A_2 w = xB_2 w + yC_2 w.$$

## Other resultant methods

- Sylvester resultants
- $u$ -resultants
- Inverse iteration, Newton-like

# More details

Which variable should the resultant method hide?

Let  $p$  and  $q$  be of degree  $(n_p, m_p, n_q, m_q)$ .

If we solve for the  $y$ -variable first,

$$B(p_y, q_y) = \sum_{i=0}^M A_i T_i(y) \in \mathbb{R}^{N \times N}, \quad \underbrace{NM = \max(n_p, n_q)(m_p + m_q)}_{\text{Size of eigenvalue problem}}.$$

If we solve for the  $x$ -variable first,

$$B(p_x, q_x) = \sum_{i=0}^M B_i T_i(x) \in \mathbb{R}^{N \times N}, \quad \underbrace{NM = \max(m_p, m_q)(n_p + n_q)}_{\text{Size of eigenvalue problem}}.$$

- Solve for  $y$ -variable first if  $\max(n_p, n_q)(m_p + m_q) \leq \max(m_p, m_q)(n_p + n_q)$ .
- **Important:** It does not change stability issues.

# More details

What is the observed computational cost?

**Cost of rootfinding is function-dependent**

Assume  $n = m_p = m_q = n_p = n_q$ .

$$O(n^6) \quad \text{vs.} \quad O(16^6 n^{-\log 4 / \log \tau})$$

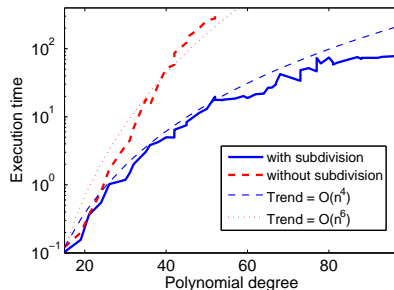
$\tau$  = average degree reduction.

$$\tau \approx 0, \quad |x||y|$$

$$\tau = \frac{1}{2}, \quad \sin(Mx)\sin(My), \quad M \gg 1$$

$$\tau = \frac{1}{\sqrt{2}}, \quad \sin(M(x - y)), \quad M \gg 1$$

$$\tau \approx 1, \quad |\sin(M(x - y))|, \quad M \gg 1$$



$$\begin{pmatrix} \sin(\omega(x + y)) \\ \cos(\omega(x - y)) \end{pmatrix} = 0, \quad 1 \leq \omega \leq 50$$

# Numerical examples

## Coordinate alignment

Solve

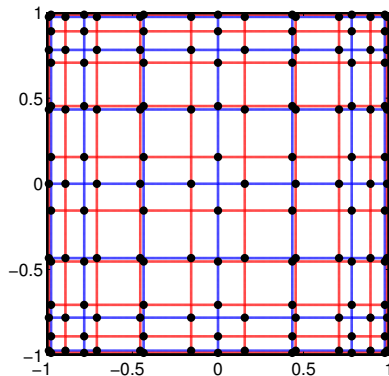
$$\begin{pmatrix} T_7(x)T_7(y)\cos(xy) \\ T_{10}(x)T_{10}(y)\cos(x^2y) \end{pmatrix} = 0.$$

Degrees are very small,

$$(m_p, n_p, m_q, n_q) = (20, 20, 24, 30),$$

but solutions aligned with grid.

$B(y)$  has **semisimple** eigenvalues with multiplicity 7 or 10. Numerically fine.



$$\text{Abs error} = 8 \times 10^{-16}$$

# Numerical examples

## Very high degree example

Find the global minimum of

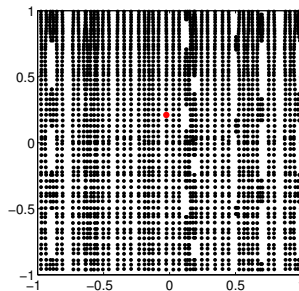
$$\begin{aligned} f(x, y) = & \left( \frac{x^2}{4} + e^{\sin(50x)} + \sin(70 \sin(x)) \right) \\ & + \left( \frac{y^2}{4} + \sin(60e^y) + \sin(\sin(80y)) \right) \\ & - \cos(10x) \sin(10y) - \sin(10x) \cos(10y). \end{aligned}$$

This example is of high degree,

$$(m_p, n_p, m_q, n_q) = (901, 625, 901, 625).$$

There are 2720 local extrema.

$$\tau \approx 0.53 \quad \Rightarrow \quad O(n^{2.2})$$



$$\text{Error} = 1.1 \times 10^{-15}$$

$$\text{Time} = 257\text{s}.$$

# Numerical examples

## Very high degree example

Find the global minimum of

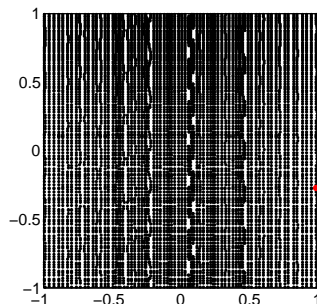
$$\begin{aligned} f(x, y) = & \left( \frac{x^2}{4} + e^{\sin(100x)} + \sin(140 \sin(x)) \right) \\ & + \left( \frac{y^2}{4} + \sin(120e^y) + \sin(\sin(160y)) \right) \\ & - \cos(20x) \sin(20y) - \sin(20x) \cos(20y). \end{aligned}$$

This example is of high degree,

(1781, 1204, 1781, 1204).

There are 9318 local extrema.

$$\tau \approx 0.5 \quad \Rightarrow \quad O(n^{2.1})$$



Time = 1300s.

# Conclusion

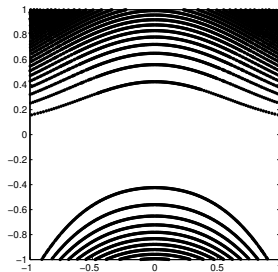
For high degree rootfinding:

- **“Polynomialization”**
- **“Act locally”**: Subdivide!
- **“Think globally”**: Stability.

For Bézout resultant:

- **Robustness from EIG**
- **Local refinement**
- **Regularization**

$$\begin{pmatrix} \text{Ai}(-13(x^2y + y^2))) \\ J_0(500x)y + xJ_1(500y) \end{pmatrix} = 0$$



$$(m_p, n_p, m_q, n_q) = (171, 120, 569, 568)$$

5932 solutions

time taken = 501s

# Thank you

Special thanks to...



Nick Trefethen



Nick Higham



Françoise Tisseur

...and to you for listening.



Y. Nakatsukasa, V. Noferini, and A. Townsend, *Computing the common zeros of two bivariate functions via Bézout resultants*, submitted, 2013.



A. Townsend, V. Noferini, and Y. Nakatsukasa, *Vector spaces of linearizations for matrix polynomials: A bivariate polynomial approach*, submitted, 2013.



A. Townsend and L. N. Trefethen, *An extension of Chebfun to two dimensions*, to appear in SISC, 2013.



# Extra slides

## Algebraic Subtleties

**Terminology:** The eigenvalues of  $B(p_y, q_y)$  satisfy

$$\det(B(p_y, q_y)) = 0.$$

If  $y_*$  is an eigenvalue then  $p(x_*, y_*) = q(x_*, y_*) = 0$  for some  $x_*$ .

**Assuming simple, isolated common zeros:**

- **Finite common zeros:**  $p(x, y_*) \neq 0$ ,  $q(x, y_*) \neq 0$  with a common finite zero, then  $y_*$  is an eigenvalue of  $B(y)$  with eigenvector  $[T_0(x_*), \dots, T_{N-1}(x_*)]^T$ .
- **Common zero at infinity:**  $p(x, y_*) \neq 0$ ,  $q(x, y_*) \neq 0$  with leading coefficient  $0T_N(x)$ .  $y_*$  eigenvalue with  $[0, \dots, 0, 1]^T$ .

If  $p(x, y_*)$  and  $q(x, y_*)$  have many common zeros  $\Rightarrow B(y)$  has a **semisimple** eigenvalue of high multiplicity.

# Extra slides

## Travelling waves

Solve

$$\begin{pmatrix} \sin(\omega x - y/\omega) + y \\ \sin(x/\omega - \omega y) - x \end{pmatrix} = 0, \quad \omega = 30.$$

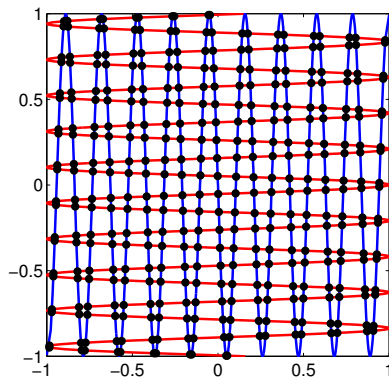
Degrees are small

$$(m_p, n_p, m_q, n_q) = (7, 63, 62, 6)$$

$$\tau \approx 0.72 \Rightarrow O(n^{4.2})$$

Subdivision in  $x$  and  $y$  independently.

Qu: Hide  $x$ - or  $y$ -variable first?



Abs error =  $1.3 \times 10^{-13}$

Time = 10.8s