

Structured matrix methods for the calculation of the roots of inexact polynomials

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1. DIFFICULTIES OF COMPUTING POLYNOMIAL ROOTS

There exist many algorithms for computing the roots of a polynomial:

- Bairstow, Graeffe, Jenkins-Traub, Laguerre, Müller, Newton, . . .

These methods yield satisfactory results if:

- The polynomial is of moderate degree
- The roots are simple and well-separated
- A good starting point in the iterative scheme is used

This heuristic has exceptions:

$$f(x) = \prod_{i=1}^{20} (x - i) = (x - 1)(x - 2) \cdots (x - 20)$$

Example 1.1 Consider the polynomial

$$x^4 - 4x^3 + 6x^2 - 4x + 1 = (x - 1)^4$$

whose root is $x = 1$ with multiplicity 4. MATLAB returns the roots

1.0002, $1.0000 + 0.0002i$, $1.0000 - 0.0002i$, 0.9998



Example 1.2 The roots of the polynomial $(x - 1)^{100}$ were computed by MATLAB.

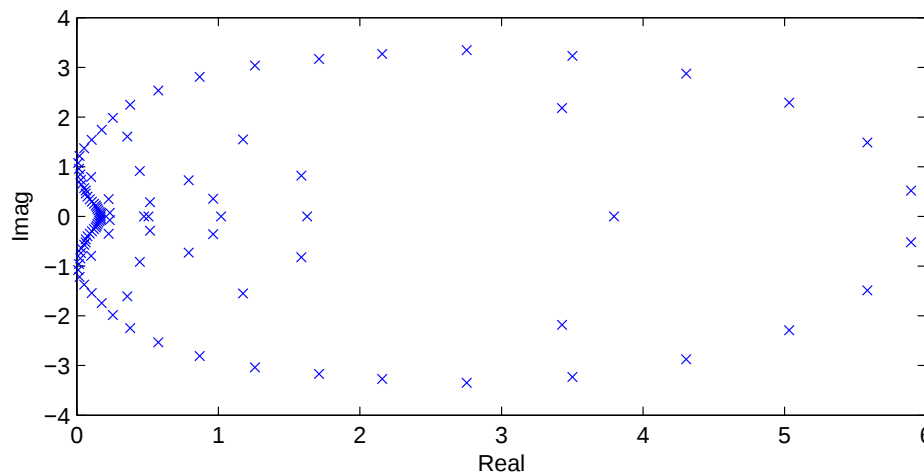


Figure 1.1: The computed roots of $(x - 1)^{100}$.



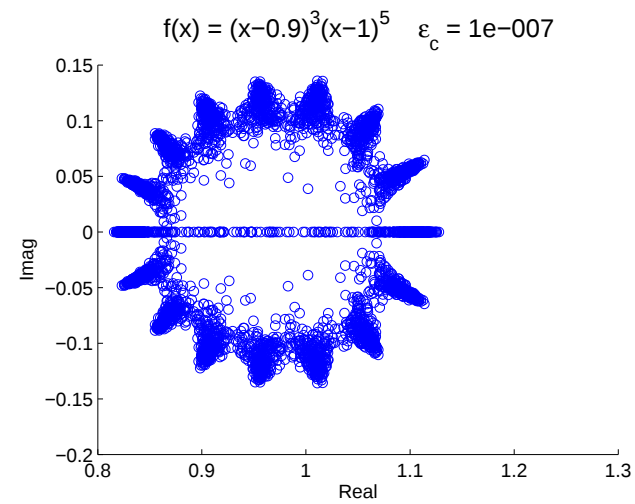
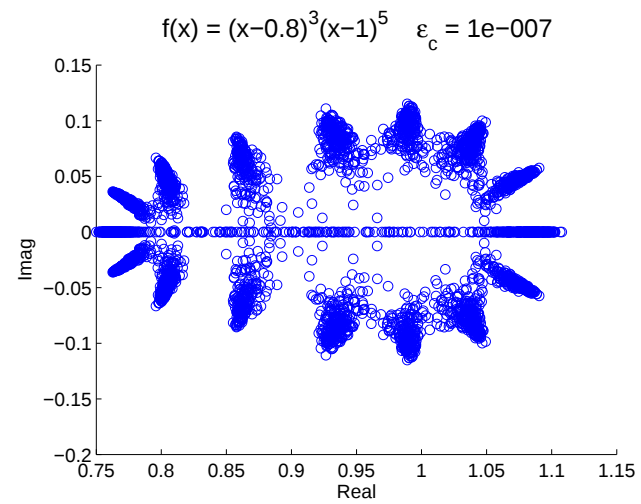
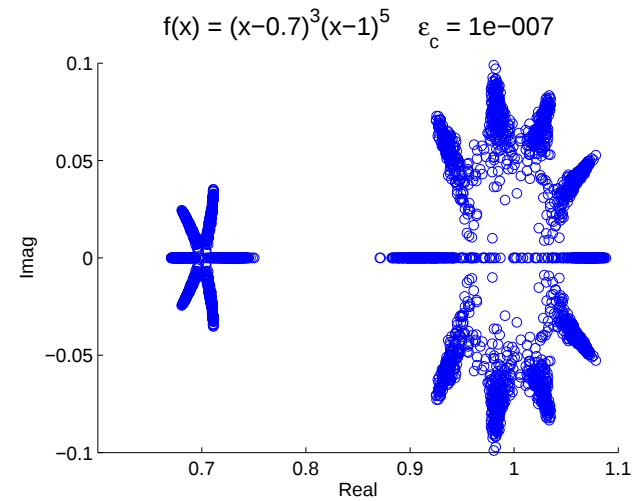
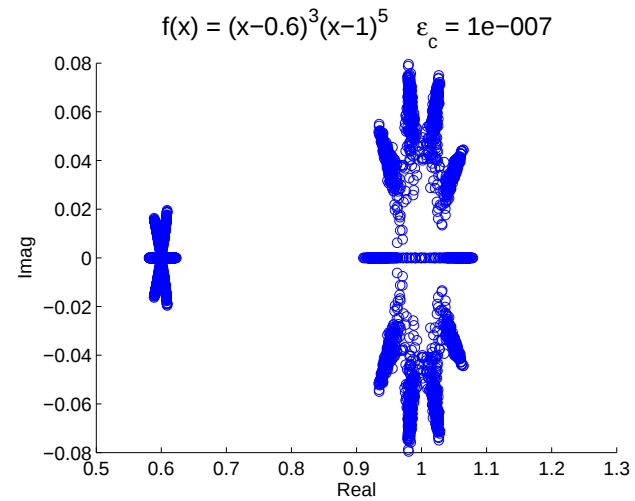


Figure 1.2: The root distribution of four perturbed polynomials.

Example 1.3

$$\frac{\text{componentwise noise amplitude}}{\text{componentwise signal amplitude}} = 10^{-8}$$

exact root	multiplicity	computed root	relative error
-6.7547000000e-1	4	-6.7547000082e-1	1.2139725913e-9
5.7335000000e+0	6	5.7335000822e+0	1.4344694923e-8
2.1747000000e+0	7	2.17469999237+0	3.5069931355e-9
-9.5568000000e+0	10	-9.5567996740e+0	3.4111255034e-8
-6.5553000000e+0	11	-6.5553001701e+0	2.5954947075e-8

- The root multiplicities were calculated correctly



Example 1.4

$$\frac{\text{componentwise noise amplitude}}{\text{componentwise signal amplitude}} = 10^{-8}$$

exact root	multiplicity	computed root	relative error
-1.1539000000e+0	4	-1.1539000316e+0	2.7389266674e-8
4.0809000000e+0	5	4.0808998890e+0	2.7198581384e-8
-2.1059000000e+0	6	-2.1058999294e+0	3.2521823798e-8
3.6683000000e+0	7	3.6683000481e+0	1.3110114060e-8
-9.6084000000e+0	13	-9.6084001121e+0	1.1664214687e-8

- The root multiplicities were calculated correctly



2. THE GEOMETRY OF ILL-CONDITIONED POLYNOMIALS

- A root x_0 of multiplicity r introduces $(r - 1)$ constraints on the coefficients.
- A monic polynomial of degree m has m degrees of freedom.
- The root x_0 lies on a manifold of dimension $(m - r + 1)$ in a space of dimension m .
- This manifold is called a **pejorative manifold** because polynomials near this manifold are ill-conditioned.
- A polynomial that lies on a pejorative manifold is well-conditioned with respect to (the structured) perturbations that keep it on the manifold, which corresponds to the situation in which the multiplicity of the roots is preserved.
- A polynomial is ill-conditioned with respect to perturbations that move it off the manifold, which corresponds to the situation in which a multiple root breaks up into a cluster of simple roots.

Example 2.1 Consider a cubic polynomial $f(x)$ with real roots x_0, x_1 and x_2

$$(x - x_0)(x - x_1)(x - x_2) = x^3 - (x_0 + x_1 + x_2)x^2 + (x_0x_1 + x_1x_2 + x_2x_0)x - x_0x_1x_2$$

- If $f(x)$ has one double root and one simple root, then $x_0 = x_1 \neq x_2$ and thus $f(x)$ can be written as

$$x^3 - (2x_1 + x_2)x^2 + (x_1^2 + 2x_1x_2)x - x_1^2x_2$$

The pejorative manifold of a cubic polynomial that has a double root is the surface defined by

$$\left(\begin{array}{ccc} -(2x_1 + x_2) & (x_1^2 + 2x_1x_2) & -x_1^2x_2 \end{array} \right) \quad x_1 \neq x_2$$

- If $f(x)$ has a triple root, then $x_0 = x_1 = x_2$ and thus $f(x)$ can be written as

$$x^3 - 3x_0x^2 + 3x_0^2x - x_0^3$$

The pejorative manifold of a cubic polynomial that has a triple root is the curve defined by

$$\begin{pmatrix} -3x_0 & 3x_0^2 & -x_0^3 \end{pmatrix}$$



Theorem 2.1 The condition number of the real root x_0 of multiplicity r of the polynomial $f(x) = (x - x_0)^r$, such that the perturbed polynomial also has a root of multiplicity r , is

$$\rho(x_0) := \frac{\Delta x_0}{\Delta f} = \frac{1}{r |x_0|} \frac{\|(x - x_0)^r\|}{\|(x - x_0)^{r-1}\|} = \frac{1}{r |x_0|} \left(\frac{\sum_{i=0}^r \binom{r}{i}^2 (x_0)^{2i}}{\sum_{i=0}^{r-1} \binom{r-1}{i}^2 (x_0)^{2i}} \right)^{\frac{1}{2}}$$

where $\|\cdot\| = \|\cdot\|_2$ and

$$\Delta f = \frac{\|\delta f\|}{\|f\|} \quad \text{and} \quad \Delta x_0 = \frac{|\delta x_0|}{|x_0|}$$

Example 2.2 The condition number $\rho(1)$ of the root $x_0 = 1$ of $(x - 1)^r$ is

$$\rho(1) = \frac{1}{r} \left(\frac{\sum_{i=0}^r \binom{r}{i}^2}{\sum_{i=0}^{r-1} \binom{r-1}{i}^2} \right)^{\frac{1}{2}}$$

This expression reduces to

$$\rho(1) = \frac{1}{r} \sqrt{\frac{\binom{2r}{r}}{\binom{2(r-1)}{r-1}}} = \frac{1}{r} \sqrt{\frac{2(2r-1)}{r}} \approx \frac{2}{r} \quad \text{for large } r$$

Compare with the componentwise and normwise condition numbers

$$\kappa_c(1) \approx \frac{|\delta x_0|}{\varepsilon_c} \quad \text{and} \quad \kappa_n(1) \approx \frac{|\delta x_0|}{\varepsilon_n}$$

- $\rho(1)$ is independent of the noise level (assumed to be small)
- $\rho(1)$ decreases as the multiplicity r of the root $x_0 = 1$ increases



3. A SIMPLE POLYNOMIAL ROOT FINDER

Let $w_i(x)$ be the product of all factors of degree i of $f(x)$

$$f(x) = w_1(x)w_2^2(x)w_3^3(x) \cdots w_{r_{\max}}^{r_{\max}}(x)$$

Perform a sequence of greatest common divisor (GCD) computations

$$\begin{aligned} q_1(x) &= \text{GCD} \left(f(x), f^{(1)}(x) \right) = w_2(x)w_3^2(x)w_4^3(x) \cdots w_{r_{\max}}^{r_{\max}-1}(x) \\ q_2(x) &= \text{GCD} \left(q_1(x), q_1^{(1)}(x) \right) = w_3(x)w_4^2(x)w_5^3(x) \cdots w_{r_{\max}}^{r_{\max}-2}(x) \\ q_3(x) &= \text{GCD} \left(q_2(x), q_2^{(1)}(x) \right) = w_4(x)w_5^2(x)w_6^3(x) \cdots w_{r_{\max}}^{r_{\max}-3}(x) \\ q_4(x) &= \text{GCD} \left(q_3(x), q_3^{(1)}(x) \right) = w_5(x)w_6^2(x)w_7^3(x) \cdots w_{r_{\max}}^{r_{\max}-4}(x) \\ &\vdots \end{aligned}$$

The sequence terminates at $q_{r_{\max}}(x)$, which is a constant.

A set of polynomials $h_i(x)$, $i = 1, \dots, r_{\max}$, is defined such that

$$h_1(x) = \frac{f(x)}{q_1(x)} = w_1(x)w_2(x)w_3(x) \cdots$$

$$h_2(x) = \frac{q_1(x)}{q_2(x)} = w_2(x)w_3(x) \cdots$$

$$h_3(x) = \frac{q_2(x)}{q_3(x)} = w_3(x) \cdots$$

$$\vdots$$

$$h_{r_{\max}}(x) = \frac{q_{r_{\max}-2}}{q_{r_{\max}-1}} = w_{r_{\max}}(x)$$

The functions, $w_1(x), w_2(x), \dots, w_{r_{\max}}(x)$, are determined from

$$w_1(x) = \frac{h_1(x)}{h_2(x)}, \quad w_2(x) = \frac{h_2(x)}{h_3(x)}, \quad \cdots, \quad w_{r_{\max}-1}(x) = \frac{h_{r_{\max}-1}(x)}{h_{r_{\max}}(x)}$$

until

$$w_{r_{\max}}(x) = h_{r_{\max}}(x)$$

The equations

$$w_1(x) = 0, \quad w_2(x) = 0, \quad \dots, \quad w_{r_{\max}}(x) = 0$$

contain only simple roots, and they yield the simple, double, triple, etc., roots of $f(x)$.

- If x_0 is a root of $w_i(x)$, then it is a root of multiplicity i of $f(x)$.



Mathematical operations performed in this root finder:

- GCD computations
- Polynomial divisions
- Solution of simple polynomial equations

Example 3.1 Calculate the roots of the polynomial

$$f(x) = x^6 - 3x^5 + 6x^3 - 3x^2 - 3x + 2$$

whose derivative is

$$f^{(1)}(x) = 6x^5 - 15x^4 + 18x^2 - 6x - 3$$

Perform a sequence of GCD computations

$$q_1(x) = \text{GCD} \left(f(x), f^{(1)}(x) \right) = x^3 - x^2 - x + 1$$

$$q_2(x) = \text{GCD} \left(q_1(x), q_1^{(1)}(x) \right) = x - 1$$

$$q_3(x) = \text{GCD} \left(q_2(x), q_2^{(1)}(x) \right) = 1$$

The maximum degree of a divisor of $f(x)$ is 3 because the sequence terminates at $q_3(x)$.

The polynomials $h_i(x)$ are:

$$h_1(x) = \frac{f(x)}{q_1(x)} = x^3 - 2x^2 - x + 2$$

$$h_2(x) = \frac{q_1(x)}{q_2(x)} = x^2 - 1$$

$$h_3(x) = \frac{q_2(x)}{q_3(x)} = x - 1$$

The polynomials $w_i(x)$ are

$$w_1(x) = \frac{h_1(x)}{h_2(x)} = x - 2$$

$$w_2(x) = \frac{h_2(x)}{h_3(x)} = x + 1$$

$$w_3(x) = h_3(x) = x - 1$$

and thus the factors of $f(x)$ are

$$f(x) = (x - 2)(x + 1)^2(x - 1)^3$$



3.1 Discussion of method

- The computation of the GCD of two polynomials is an ill-posed problem because it is not a continuous function of their coefficients:
 - The polynomials $f(x)$ and $g(x)$ may have a non-constant GCD, but the perturbed polynomials $f(x) + \delta f(x)$ and $g(x) + \delta g(x)$ may be coprime.
- The determination of the degree of the GCD of two polynomials reduces to the determination of the rank of a resultant matrix, but the rank of a matrix is not defined in a floating point environment.
- Polynomial division is an ill-posed problem:

Even if $\frac{f(x)}{g(x)}$ is a polynomial,

$\frac{f(x) + \delta f(x)}{g(x) + \delta g(x)}$ is a rational function for arbitrary $\delta f(x)$ and $\delta g(x)$

4. APPROXIMATE GREATEST COMMON DIVISORS

If $f(x)$ is exact and all computations are performed in a symbolic environment, the GCD of $f(x)$ and its derivative $f^{(1)}(x)$ can be computed by the **Sylvester resultant matrix** $S(f, f^{(1)})$.

The polynomial $f(x)$ is rarely known exactly, and so the given data is

$$\tilde{f}(x) = f(x) + \delta f(x)$$

and $\tilde{f}(x)$ and $\tilde{f}^{(1)}(x)$ are (with probability almost 1) coprime.

- The polynomials $\tilde{f}(x)$ and $\tilde{f}^{(1)}(x)$ have an **approximate greatest common divisor (AGCD)**.
- Use the method of **structured total least norm** applied to $S(\tilde{f}, \tilde{f}^{(1)})$ to compute the smallest perturbation of $S(\tilde{f}, \tilde{f}^{(1)})$ such that its perturbed form is singular, which implies that the perturbed form $\tilde{f}(x)$ of $f(x)$ has a multiple root.

4.1 The Sylvester resultant matrix

The product of two polynomials is equal to the convolution of their coefficients:

$$\begin{bmatrix} r_0 \\ r_1 \\ \vdots \\ r_{m+n-1} \\ r_{m+n} \end{bmatrix} = \begin{bmatrix} p_0 & & & & \\ p_1 & p_0 & & & \\ \vdots & p_1 & \ddots & & \\ p_m & \vdots & \ddots & p_0 & \\ & p_m & \ddots & p_1 & \\ & & \ddots & \vdots & \\ & & & p_m & \end{bmatrix} \begin{bmatrix} q_0 \\ q_1 \\ \vdots \\ q_{n-1} \\ q_n \end{bmatrix}$$

$$\mathbf{r} = C_{n+1}(p)\mathbf{q} = \mathbf{p} \otimes \mathbf{q}$$

$$\mathbf{r} \in \mathbb{R}^{m+n+1}, \mathbf{p} \in \mathbb{R}^{m+1}, \mathbf{q} \in \mathbb{R}^{n+1} \text{ and } C_{n+1}(p) \in \mathbb{R}^{(m+n+1) \times (n+1)}$$

Let:

- $d_k(y)$ be a common divisor of degree k of the exact polynomials $f(y)$ and $f^{(1)}(y)$
- The degree of the GCD of $f(y)$ and $f^{(1)}(y)$ be $\hat{d}$
- $u_k(y)$ and $v_k(y)$ be the quotient polynomials

$$f(y) = u_k(y)d_k(y) \quad \text{and} \quad f^{(1)}(y) = v_k(y)d_k(y)$$

Thus

$$f(y)v_k(y) - f^{(1)}(y)u_k(y) = 0 \quad \Leftrightarrow \quad C_{m-k}(f)\mathbf{v}_k - C_{m-k+1}(f^{(1)})\mathbf{u}_k = 0$$

where

$$\begin{aligned} C_{m-k}(f) &\in \mathbb{R}^{(2m-k) \times (m-k)} & \text{and} \quad \mathbf{v}_k &\in \mathbb{R}^{m-k} \\ C_{m-k+1}(f^{(1)}) &\in \mathbb{R}^{(2m-k) \times (m-k+1)} & \text{and} \quad \mathbf{u}_k &\in \mathbb{R}^{m-k+1} \end{aligned}$$

$$\begin{bmatrix} C_{m-k}(f) & C_{m-k+1}(f^{(1)}) \end{bmatrix} \begin{bmatrix} \mathbf{v}_k \\ -\mathbf{u}_k \end{bmatrix} = S_k(f, f^{(1)}) \begin{bmatrix} \mathbf{v}_k \\ -\mathbf{u}_k \end{bmatrix} = 0$$

- $S_k(f, f^{(1)}) \in \mathbb{R}^{(2m-k) \times (2m-2k+1)}$ and it is rank deficient
- The nullspace vectors yield the coefficients of the quotient polynomials
- Since the degree of the GCD of $f(y)$ and $f^{(1)}(y)$ is $\hat{d}$, these polynomials possess common divisors of degrees $1, 2, \dots, \hat{d}$, but not a divisor of degree $\hat{d} + 1$:

$$\text{rank } S_k(f, f^{(1)}) < 2m - 2k + 1, \quad k = 1, \dots, \hat{d}$$

$$\text{rank } S_k(f, f^{(1)}) = 2m - 2k + 1, \quad k = \hat{d} + 1, \dots, m - 1$$

Calculating the degree of the GCD reduces to estimating the rank of a matrix

Example 4.1 Consider $S_k(f, f^{(1)})$, for $k = 1, 2, 3$, for

$$f(x) = (x-1)^2(x-2)(x-3) = x^4 - 7x^3 + 17x^2 - 17x + 6$$

$$f^{(1)}(x) = 4x^3 - 21x^2 + 34x - 17$$

Hence $S_1(f, f^{(1)}) = S(f, f^{(1)})$ is equal to

$$\begin{bmatrix} 1 & 0 & 0 & 4 & 0 & 0 & 0 \\ -7 & 1 & 0 & -21 & 4 & 0 & 0 \\ 17 & -7 & 1 & 34 & -21 & 4 & 0 \\ -17 & 17 & -7 & -17 & 34 & -21 & 4 \\ 6 & -17 & 17 & 0 & -17 & 34 & -21 \\ 0 & 6 & -17 & 0 & 0 & -17 & 34 \\ 0 & 0 & 6 & 0 & 0 & 0 & -17 \end{bmatrix}$$

and this matrix has a unit loss of rank.

The subresultant matrix $S_2(f, f^{(1)})$ is

$$S_2(f, f^{(1)}) = \begin{bmatrix} 1 & 0 & 4 & 0 & 0 \\ -7 & 1 & -21 & 4 & 0 \\ 17 & -7 & 34 & -21 & 4 \\ -17 & 17 & -17 & 34 & -21 \\ 6 & -17 & 0 & -17 & 34 \\ 0 & 6 & 0 & 0 & -17 \end{bmatrix}$$

and this matrix has full column rank.

The subresultant matrix $S_3(f, f^{(1)})$ is

$$S_3(f, f^{(1)}) = \begin{bmatrix} 4 & 0 & 1 \\ -21 & 4 & -7 \\ 34 & -21 & 17 \\ -17 & 34 & -17 \\ 0 & -17 & 6 \end{bmatrix}$$

and this matrix has full column rank.

It follows that the first rank deficient matrix in the sequence

$$S_3(f, f^{(1)}), S_2(f, f^{(1)}), S_1(f, f^{(1)})$$

is $S_1(f, f^{(1)})$, and thus the degree of the GCD of $f(x)$ and $f^{(1)}(x)$ is one. $\square$

4.2 Pre-processing operations for the computation of an AGCD

The computation of an AGCD of $f(x)$ and $f^{(1)}(x)$ requires that two pre-processing operations be performed:

- $f(x)$ and $f^{(1)}(x)$ must be normalised to balance the Sylvester matrix
- An AGCD of $f(x)$ and $f^{(1)}(x)$ is equal to, up to a scalar multiplier, an AGCD of $f(x)$ and $\alpha f^{(1)}(x)$, where α is an arbitrary non-zero constant.

$$\text{GCD} \left(f, f^{(1)} \right) \sim \text{GCD} \left(f, \alpha f^{(1)} \right), \quad \alpha \neq 0$$

- The resultant matrix $S(f, \alpha f^{(1)})$ should be used when it is desired to compute an AGCD of $f(x)$ and $f^{(1)}(x)$
- How is the optimal value of α computed?

1. Normalisation: Define $f(x)$ and $g(x)$ as

$$f(x) = \sum_{i=0}^m \bar{a}_i x^{m-i}, \quad \bar{a}_i = \frac{a_i}{\left(\prod_{j=0}^m |a_j|\right)^{\frac{1}{m+1}}}$$
$$g(x) = \sum_{i=0}^{m-1} \bar{b}_i x^{m-1-i}, \quad \bar{b}_i = \frac{(m-i)\bar{a}_i}{\left(\prod_{j=0}^{m-1} |(m-j)\bar{a}_j|\right)^{\frac{1}{m}}}$$

Note: $g(x)$ is proportional to $f^{(1)}(x)$

2. The optimal value of α : Use linear programming to calculate α_0 , the optimal value of α

4.3 The coprime polynomials and the degree of an AGCD

Recall that

$$S_k(f, g) \begin{bmatrix} \mathbf{v}_k \\ -\mathbf{u}_k \end{bmatrix} = 0$$

- $S_k(f, g) \in \mathbb{R}^{(2m-k) \times (2m-2k+1)}$ and it is rank deficient
- The nullspace vectors yield the coefficients of the quotient polynomials
- If the degree of an GCD of $f(x)$ and $g(x)$ is d , then

$$\text{rank } S_k(f, g) < 2m - 2k + 1, \quad k = 1, \dots, d$$

$$\text{rank } S_k(f, g) = 2m - 2k + 1, \quad k = d + 1, \dots, m - 1$$

Use the same criterion for the calculation of the degree of an AGCD

- The degree d of an AGCD of $f(x)$ and $g(x)$ is equal to the largest value of k , $k = 1, \dots, m - 1$, such that $S_k(f, g)$ is numerically singular:
 - The SVD of $S_k(f, g)$ cannot be used because $f(x)$ and $g(x)$ are inexact and therefore, with high probability, coprime
 - The property

$$\text{numerical rank } S_k(f, \alpha_0 g) = \text{numerical rank } S_k \left(g, \frac{f}{\alpha_0} \right)$$

enables a criterion for the calculation of d to be developed

- $S_d(f, \alpha_0 g)$ is numerically rank deficient by one and estimates of the coprime polynomials can be calculated from its nullspace

$$S_d(f, \alpha_0 g) \begin{bmatrix} \mathbf{v}_d \\ -\mathbf{u}_d \end{bmatrix} \approx 0$$

5. STRUCTURED TOTAL LEAST NORM

Recall that $S_d(f, \alpha_0 g)$ is numerically rank deficient by one and

$$S_d(f, \alpha_0 g) \begin{bmatrix} \mathbf{v}_d \\ -\mathbf{u}_d \end{bmatrix} \approx 0$$

If this equation is satisfied exactly, then

- The coprime polynomials are defined by the null space of $S_d(f, \alpha_0 g)$
- $f(y)$ and $g(y)$ have a non-constant common divisor:
 - $f(y)$ has a multiple root
 - $g(y)$ has been moved to a pejorative manifold

The calculation of d determines the column q of $S_d(f, \alpha_0 g)$ such that if

$$S_d(f, \alpha_0 g) = \begin{bmatrix} c_1 & c_1 & \cdots & c_{q-1} & c_q & c_{q+1} & \cdots & c_{2m-2k+1} \end{bmatrix}$$

then the approximate homogeneous equation

$$S_d(f, \alpha_0 g) \begin{bmatrix} \mathbf{v}_d \\ -\mathbf{u}_d \end{bmatrix} \approx 0$$

can be transformed into an approximate linear algebraic equation

$$\begin{bmatrix} c_1 & c_1 & \cdots & c_{q-1} & c_{q+1} & \cdots & c_{2m-2d+1} \end{bmatrix} x \approx c_q$$

or

$$A_d x \approx c_q, \quad A_d \in \mathbb{R}^{(2m-d) \times (2m-2d)}, \quad c_q \in \mathbb{R}^{2m-d}$$

$$x = \begin{bmatrix} x_1 \\ \vdots \\ x_{q-1} \\ x_{q+1} \\ \vdots \\ x_{2m-2d} \end{bmatrix} \in \mathbb{R}^{2m-2d}, \quad x_q = -1, \quad \begin{bmatrix} x_1 \\ \vdots \\ x_{q-1} \\ -1 \\ x_{q+1} \\ \vdots \\ x_{2m-2d} \end{bmatrix} = \begin{bmatrix} \mathbf{v}_d \\ -\mathbf{u}_d \end{bmatrix}$$

Recall:

- The degree of an AGCD of the given inexact polynomials $f(y)$ and $g(y)$ is d
- The matrix A_d and vector c_q are functions of the coefficients of $f(y)$ and $g(y)$.
-

$$A_d x \approx c_q$$

is an approximate equation because its arguments are the coefficients of inexact polynomials

Use structured total least norm to solve this approximate equation

Given the inexact polynomials $f(y)$ and $g(y)$, which are assumed to be coprime, calculate the smallest perturbations that must be added to their coefficients such that the perturbed forms of $f(y)$ and $g(y)$ have a non-constant GCD.

Aim:

Compute the Sylvester matrix $S(\delta f, \alpha_0 \delta g)$, such that

$$\|\delta f\|^2 + \|\delta g\|^2$$

is minimised, where

$$S_d(f + \delta f, \alpha_0(g + \delta g)) = S_d(f, \alpha_0 g) + S_d(\delta f, \alpha_0 \delta g)$$

is rank deficient.

The approximate under-determined equation

$$A_d x \approx c_q$$

is corrected by considering the equation

$$(A_d(\alpha_0) + E_d(\alpha_0, z)) x = c_q(\alpha_0) + h_q(\alpha_0, z)$$

which is non-linear in x and z .

- A_d and E_d have the same structure, and c_d and h_d have the same structure:
 - Use the method of structured total least norm
- The initial vector of perturbations is $z = 0$
- Solve this non-linear under-determined equation subject to the constraint that $\|z\|^2$ is minimised
- This leads to a least squares equality problem

The Sylvester matrix $S_d(f, \alpha_0 g)$ is

$$\begin{bmatrix} \bar{a} & & & \alpha_0 \bar{b} & & \\ \bar{a}_1 & \ddots & & \alpha_0 \bar{b}_1 & \ddots & \\ \vdots & \ddots & \bar{a} & \vdots & \ddots & \alpha_0 \bar{b} \\ \bar{a}_{m-1} & \ddots & \bar{a}_1 & \alpha_0 \bar{b}_{n-1} & \ddots & \alpha_0 \bar{b}_1 \\ \bar{a}_m & \ddots & \vdots & \alpha_0 \bar{b}_n & \ddots & \vdots \\ & \ddots & \bar{a}_{m-1} & & \ddots & \alpha_0 \bar{b}_{n-1} \\ & & \bar{a}_m & & & \alpha_0 \bar{b}_n \end{bmatrix}$$

If the perturbations of the coefficients of $f(y)$ and $\alpha_0 g(y)$ are

$$z_i, i = 0, \dots, m \quad \text{and} \quad \alpha_0 z_{m+1+i}, i = 0, \dots, n$$

respectively, then $E_d(\alpha_0, z)$ is equal to

$$\begin{bmatrix} z_0 & & & \alpha_0 z_{m+1} & & \\ z_1 & \ddots & & \alpha_0 z_{m+2} & \ddots & \\ \vdots & \ddots & & \vdots & \ddots & \alpha_0 z_{m+1} \\ z_{m-1} & \ddots & z_1 & \alpha_0 z_{m+n} & \ddots & \alpha_0 z_{m+2} \\ z_m & \ddots & \vdots & \alpha_0 z_{m+n+1} & \ddots & \vdots \\ & \ddots & z_{m-1} & & \ddots & \alpha_0 z_{m+n} \\ & & z_m & & & \alpha_0 z_{m+n+1} \end{bmatrix}$$

- Perform these AGCD computations repeatedly in order to determine the multiplicities of the roots
 - These calculations correspond to the identification of the pejorative manifold on which the theoretically exact polynomial lies

The other stages in the algorithm:

- Use the method of least squares to perform the polynomial division
- Recall it is necessary to solve the polynomial equations

$$w_1(x) = 0, \quad w_2(x) = 0, \quad \dots, \quad w_{r_{\max}}(x) = 0$$

- Calculate the roots of the polynomial by solving these equations

6. EXAMPLES

Example 6.1

$$\frac{\text{componentwise noise amplitude}}{\text{componentwise signal amplitude}} = 10^{-6}$$

exact root	multiplicity	computed root	relative error
8.1031100000e+0	2	8.1031311463e+0	3.8437554630e-6
3.5078000000e+0	8	3.5077983383e+0	4.7372726251e-7
-6.3060000000e-1	8	-6.3060013449e-1	2.1327857935e-7
-5.8211000000e+0	9	-5.8210973315e+0	4.5841110328e-7

- The root multiplicities were calculated correctly



Example 6.2

$$\frac{\text{componentwise noise amplitude}}{\text{componentwise signal amplitude}} = 10^{-8}$$

exact root	multiplicity	computed root	relative error
-7.3132000000e+0	1	-7.3131318042e+0	9.3250329595e-6
9.0183000000e+0	2	9.0182738917e+0	2.8950322472e-6
4.4470000000e+0	3	4.4469987289e+0	2.8582679636e-7
6.6374000000e+0	4	6.6374090279e+0	1.3601587262e-6
-1.9984000000e+0	4	-1.9984000974e+0	4.8743537340e-8
-8.7907000000e+0	6	-8.7907151653e+0	1.7251509523e-6

- The root multiplicities were calculated correctly

